

AD-A110 322

DEFT LABS INC EAST SYRACUSE NY

F/G 20/1

ACOUSTO-OPTIC TECHNOLOGY FOR TOPOGRAPHIC FEATURE EXTRACTION AND--ETC(U)

MAR 81 A L MOYER

DAAK70-79-C-0160

0102-A002

ETL-0256

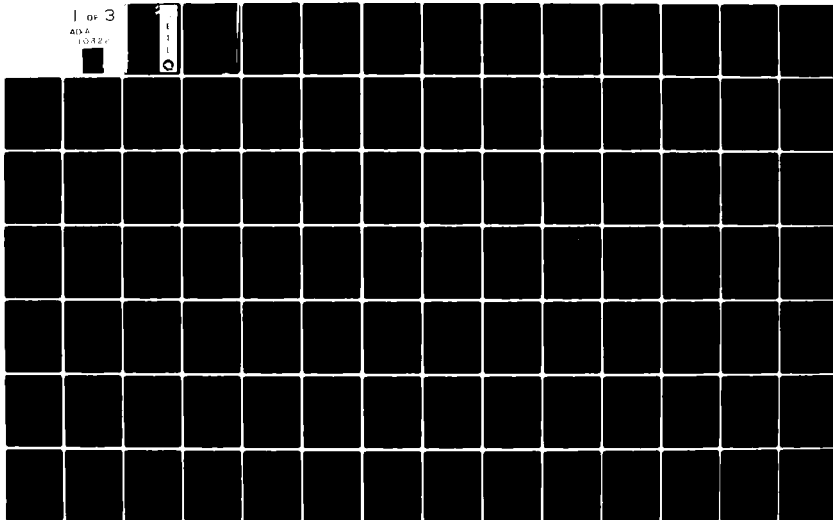
NL

UNCLASSIFIED

1 of 3

AD-A110 322

10/12/00



1.0

1.8 2.5

2.2

1.1

2.0

1.8

1.25

1.4

1.6

U.S. GOVERNMENT PRINTING OFFICE: 1963

10-40000

AD A110322

ETL-0256

LEVEL II

Acousto-optic technology for  
topographic feature extraction  
and image analysis

Alan L. Moyer

Deft Laboratories, Inc.  
7 Adler Drive  
East Syracuse, NY 13057

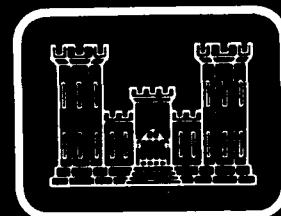
DTIC  
ELECTE  
FEB 0 2 1982  
S D E

MARCH 1981

APPROVED FOR PUBLIC RELEASE: DISTRIBUTION UNLIMITED

DTIC FILE COPY

U.S. ARMY CORPS OF ENGINEERS  
ENGINEER TOPOGRAPHIC LABORATORIES  
FORT BELVOIR, VIRGINIA 22060



E

T

L



Destroy this report when no longer needed.  
Do not return it to the originator.

---

The findings in this report are not to be construed as an official  
Department of the Army position unless so designated by other  
authorized documents.

---

The citation in this report of trade names of commercially available  
products does not constitute official endorsement or approval of the  
use of such products.



Unclassified

SECURITY CLASSIFICATION OF THIS PAGE (When Data Entered)

REPORT DOCUMENTATION PAGE		READ INSTRUCTIONS BEFORE COMPLETING FORM
1. REPORT NUMBER ETL-0256	2. GOVT ACCESSION NO. AD-A420 322	3. RECIPIENT'S CATALOG NUMBER
4. TITLE (and Subtitle) ACOUSTO-OPTIC TECHNOLOGY FOR TOPOGRAPHIC FEATURE EXTRACTION AND IMAGE ANALYSIS		5. TYPE OF REPORT & PERIOD COVERED Final Technical Report Oct. 1979 thru Feb. 1981
7. AUTHOR(s) Alan L. Moyer		6. PERFORMING ORG. REPORT NUMBER 0102-A002
9. PERFORMING ORGANIZATION NAME AND ADDRESS Deft Laboratories, Inc. 7 Adler Drive East Syracuse, NY 13057		8. CONTRACT OR GRANT NUMBER(s) DAAK70-79-C-0160
11. CONTROLLING OFFICE NAME AND ADDRESS U.S. Army Engineer Topographic Laboratories Fort Belvoir, VA 22060		10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS
14. MONITORING AGENCY NAME & ADDRESS (if different from Controlling Office)		12. REPORT DATE March 1981
		13. NUMBER OF PAGES 205
		15. SECURITY CLASS. (of this report) Unclassified
		15a. DECLASSIFICATION/DOWNGRADING SCHEDULE
16. DISTRIBUTION STATEMENT (of this Report)  Approved for public release; distribution unlimited		
17. DISTRIBUTION STATEMENT (of the abstract entered in Block 20, if different from Report)		
18. SUPPLEMENTARY NOTES		
19. KEY WORDS (Continue on reverse side if necessary and identify by block number)  Image Processing, Acousto-Optics, Algorithms, Feature Extraction		
20. ABSTRACT (Continue on reverse side if necessary and identify by block number)  This report contains all findings of the acousto-optic technology study for feature extraction conducted by Deft Laboratories Inc. for the U.S. Army Engineer Topographic Laboratories. The objective of this program was to develop, analyze and evaluate theoretical concepts and strategies for topographic feature extraction and image analysis using acousto-optic (A-O) technology.		

DD FORM 1 JAN 73 1473 EDITION OF 1 NOV 65 IS OBSOLETE

Unclassified

SECURITY CLASSIFICATION OF THIS PAGE (When Data Entered)

412038

## 20. Abstract (cont'd.)

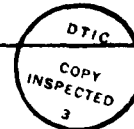
A conclusion of this study was that A-0 devices are potentially capable of implementing the feature extraction prefilter function very efficiently. Since the prefilter is the most computational intensive portion of the feature extraction process this is a significant result. The best application of A-0 devices is in the implementation of transform-based prefilter algorithms. Under this contract transform-based algorithms were identified and developed which are invariant to feature translation, rotation and scale. This invariance is highly desirable since it reduces the number of distinct feature signatures which must be processed by the decision processor.

Some preliminary experiments were conducted using the Fourier-based algorithms, test images and A-0 device which was a Deft sensor. This combination of algorithms and sensor was able to distinguish between three test patterns which were presented in arbitrary orientation and scale. The success rate was 80%. In spite of these promising results, present Deft sensors are not capable of distinguishing realistic features in aerial photographs. New Deft sensors, presently under development, are described which will significantly improve the capability of this sensor in feature extraction applications.

# PREFACE

This report contains all findings of the acousto-optic technology study for feature extraction conducted by Deft Laboratories Inc. for the U.S. Army Engineer Topographic Laboratories. The work reported here was funded under Contract DAAK 70-79-C-0160. The work was conducted during the period of October 1979 through February 1981. The Contracting Office's Technical Representative was Mr. Joseph F. Hannigan.

<b>Accession For</b>	
NTIS GRA&I	<input checked="" type="checkbox"/>
DTIC TAB	<input type="checkbox"/>
Unannounced	<input type="checkbox"/>
Justification	
By	
Distribution	
Availability Codes	
Dist	Special
A	



## SUMMARY

The objective of this program was to develop, analyze and evaluate theoretical concepts and strategies for topographic feature extraction and image analysis using acousto-optic (A-O) technology.

A conclusion of this study was that A-O devices are potentially capable of implementing the feature extraction prefilter function very efficiently. Since the prefilter is the most computational intensive portion of the feature extraction process this is a significant result. The best application of A-O devices is in the implementation of transform-based prefilter algorithms. Under this contract transform-based algorithms were identified and developed which are invariant to feature translation, rotation and scale. This invariance is highly desirable since it reduces the number of distinct feature signatures which must be processed by the decision processor.

Some preliminary experiments were conducted using the Fourier-based algorithms, test images and an A-O device which was a Deft sensor. This combination of algorithms and sensor was able to distinguish between three test patterns which were presented in arbitrary orientation and scale. The success rate was 80%. In spite of these promising results, present Deft sensors are not capable of distinguishing realistic features in aerial photographs. New Deft sensors are presently under development which will significantly improve the capability of this sensor in feature extraction applications.

# TABLE OF CONTENTS

	<u>Page</u>
Preface	iii
Summary	iv
List of Figures	vi-vii
List of Tables	viii
I. Introduction	1
A. Project Overview	1
B. General Feature Extraction System Model	2
II. Acousto-Optic Devices	5
A. Introduction	5
B. The Deft Sensor	5
C. Elastobirefringent Light Valve	10
D. Thomson - CSF Sensor	13
III. Feature Extraction Techniques for Acousto-Optic Technology	17
A. Introduction	17
B. General Concepts	17
C. Transform Methods	24
1. Hadamard Transform	25
2. Matched Filtering	32
3. Method of Invariant Fourier Signatures	38
4. Method of Invariant Moment Signatures	53
D. Summary	80
IV. Feature Extraction Experiments with Deft Sensors	83
A. Introduction	83
B. Feature Extraction Experimental Computer Programs	84
1. Method of Invariant Moment Signatures	84
2. Method of Invariant Fourier Signatures	91
C. Feature Extraction Experiments	98
V. Acousto-Optic Sensor Capabilities: Present and Projected	113
A. Introduction	113
B. Elastobirefringent Light Valve	113
C. Deft Sensor	114
VI. Summary and Conclusions	123
VII. References	125
Appendix A - Method of Invariant Fourier Signatures Assembly Code Listing	
Appendix B - Method of Invariant Moments Assembly Code Listing	
Appendix C - Fortran Code Listings	

# LIST OF FIGURES

<u>FIGURE NO.</u>		<u>PAGE</u>
1	Pattern Recognition System Model	4
2	Deft Sensor Construction	6
3	Periodic Extension of $g(x) = x$	9
4	Elastobirefringent Light Valve: Experimental Set-Up	11
5	Thomson - CSF Sensor	13
6	Transducer Instantaneous Frequencies	14
7	Raster Scan of Transform Plane	15
8	Radial-Scanned Transform	15
9	Image Scanning: No Overlap	20
10	Image Scanning: 50% Overlap	20
11	Overlap Geometry	21
12	First Eight Walsh Functions	25
13	Modulation of Sinusoid by a Walsh Function	27
14	Walsh/Fourier Bandwidth Relationship	31
15	Matched Filter: A-0 Implementation	35
16	FF* Matched Filter: A-0 Implementation	37
17	Integration Contours	40
18	$(\omega_\rho, \omega_\theta)$ - Space and $F\{P\phi\}$ Line	45
19	$(\omega_x, \omega_y)$ - Space Contours	47
20	$(\rho, \theta)$ - Space Contours	47
21	Invariant Fourier Signatures: Implementation	52
22	Equivalent Image Ellipse	55
23	Transform Sampling: Finite Differences	59
24	Periodic Extensions of $x^p$	62
25	Fourier Transform Sample Spacing for Moment Computation	71
26	$e_{p,n}(x)$ for $p = 1$ and $n/4=1,2$ and $4$	73
27	Implementation: Method of Moments	80
28	Flow Diagram: Invariant Moment Signatures	85

LIST OF FIGURES - Cont'd.

<u>FIGURE NO.</u>		<u>PAGE</u>
29	Moments for Feature Extraction Sample Run	88
30	Flow Diagram: Invariant Fourier Signatures	92
31	Flow Diagram Detail: Projection Computation	93
32	Invariant Fourier Signatures: Sample Run	97
33	Test Reference	102
34	Experimental Set-up	103
35	Deft Transform of "Crossroad" Feature	116
36	Deft Transform of "Road" Feature	117
37	Deft Transform of Small Circle Feature	118

# LIST OF TABLES

<u>TABLE NO.</u>		<u>PAGE</u>
1	$S(n,f)$ vs. $n$	29
2	$C(n,f)$ vs. $n$	30
3	Maximum $ e_{p,n}(x) $ vs. $p$ and $n/n_p$	74
4	Maximum $ e_{p,n}(x) $ vs. $p = n/n_p$ or $p = n/n_p - 1$	76
5	Program Coefficients for $p,q$ Even	89
6	Program Coefficients for $p,q$ Odd	90
7	Feature Extraction Experimental Results: Sensor #1, 32 Radial Lines	106
8	Feature Extraction Experimental Results: Sensor #1, 32 Circles	107
9	Feature Extraction Experimental Results: Sensor #2, 32 Radial Lines	108
10	Feature Extraction Experimental Results: Sensor #2, 32 Circles	108
11	Reference Pattern Correlations	110



## I. INTRODUCTION

### A. Project Overview

The objective of this program was to develop, analyze and evaluate theoretical concepts and strategies for topographic feature extraction and image analysis using acousto-optic (A-O) technology. In addition, functional block diagrams were to be prepared for the most promising concepts and strategies for the purpose of identifying essential and/or critical A-O elements and components.

During the execution of the program it became evident that the evaluation of theoretical concepts was sufficiently advanced so that additional effort, not required in the original contract, could be undertaken. This effort involved programming a group of promising feature extraction algorithms and performing experiments using an A-O device. The device used in these experiments was a Deft sensor.

The tasks which were executed during this program are listed below.

1. The open literature was searched in the areas of image processing, feature extraction and acousto-optic technology.
2. A survey was conducted and determination made of device capability and limitations of present A-O technology.
3. Theoretical concepts and strategies were analyzed and evaluated for topographic feature extraction and for implementation using A-O technology.
4. The most promising concepts and strategies were selected for further evaluation.
5. For these strategies algorithms were developed and programmed for a microprocessor-based experimental system. Experiments were conducted using this system and a Deft sensor. These experiments involved feature recognition using a small set of test images.
6. Functional block diagrams were developed for the most

promising strategies. Essential and/or critical A-0 elements were identified.

7. For these A-0 elements state-of-the-art capabilities were compared with required capabilities. Required improvements were determined as well as indication of the probability of obtaining such improvements.

This report details all findings of this study program.

#### B. General Feature Extraction System Model

The term "feature extraction" comes under the heading of pattern recognition. Large volumes of material have been written under this heading. The specific problem addressed by this study is how can A-0 devices be used to find objects or "features" in aerial photographs? The emphasis, then, is on what functions A-0 devices can perform and how these functions can be used to achieve the stated objective. In this context, only certain portions of the literature under pattern recognition has bearing on the problem to be solved.

It was felt at the beginning of this study that the only way to achieve meaningful results was to determine a model for a feature extraction system which has the general functional capability required. From this general model, more specific systems could be developed which utilize functions which can be efficiently performed by A-0 devices. Concepts and strategies could then be chosen from the literature under pattern recognition which fit into these specific system models.

In order to provide a frame of reference, feature extraction will be considered in the context of the pattern-recognition system model shown in Figure 1. The transducer transforms the information content of the input (photolight intensity pattern) into a format suitable for further processing (electrical signal). The amount of information available in the photo image is enormous. A single video image may contain  $10^4$  -  $10^8$  bits of

information, while a photographic image can contain orders of magnitude more. However, the amount of information required to make a decision may be just a few tens of numbers. The purpose of the preprocessor is to reduce the information content of the image to a more manageable size. This process has also been termed filtering, prefiltering, feature or measurement extraction or dimensionality reduction. The output of the preprocessor is then passed to the decision processor where this information is used to classify the photo image. The decision processor may also control the transducer and preprocessor via feedback. In this manner the preprocessor function may be changed during the decision process. In like manner the transducer may be commanded to look at a different portion of the photo or perhaps change scale.

Figure 1 is a functional block diagram. These functions may be distributed over hardware subsystems in a number of ways. For example, the transducer might be a vidicon or CCD imager. Its function then is primarily bandwidth reduction and format conversion. The preprocessor and decision processor might then be implemented in software in a digital computer.

Another alternative configuration is considered in some detail in this report. In this configuration the transducer and preprocessor functions are performed in an acousto-optic device. The decision process is then carried out in a digital processor. In an all digital implementation the preprocessor function requires the bulk of the processing time since a large data base (image) must be operated on. The potential advantage of the acousto-optic implementation is that this time-consuming preprocessing function can be performed in whole or in part in the sensor itself.

A third configuration should be mentioned. In this configuration the transducer and preprocessor functions are performed by an optical processor. Many of the prefilter functions which can be computed in an optical processor can also be computed by acousto-

optic devices. As a result, the large literature which deals with applications of optical processors can be utilized to develop preprocessing techniques for acousto-optic devices. The development of both optical processors and acousto-optic devices is currently receiving support because of their potential application in such areas as image preprocessing. At this time, optical processors are in a more advanced state of development than acousto-optic devices. However, acousto-optic devices offer the potential advantages of being cheaper, more rugged and not requiring precise optical alignment when compared to optical processors. In certain applications they may also prove to be more flexible since the function performed can be controlled electrically.

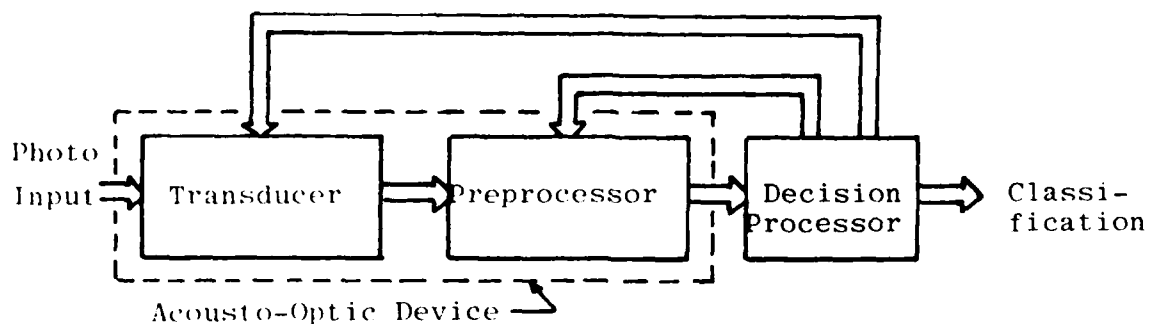


Figure 1 - Pattern Recognition System Model<sup>1</sup>

## II. ACOUSTO-OPTIC DEVICES

### A. Introduction

In this section three acousto-optic devices are briefly described. Both physical and functional descriptions are provided. From these descriptions it will then be possible to determine which feature extraction algorithms are best suited for implementation using one or more of these devices. More detailed descriptions can be found in the references. The devices to be described are the following: the Deft sensor manufactured by Deft Laboratories Inc., an elastobirefringent light valve and a device developed by Thomson-CSF, France. Both the Deft sensor and the elastobirefringent light valve were developed by Drs. Kowel and Kornreich of Deft Laboratories. As a result we are more familiar with the present limitations and future potential of these devices. Section V is devoted to a discussion of this topic.

### B. The Deft Sensor

The Deft sensor is a solid state device which utilizes the interaction of surface acoustic waves (SAW), a photoconducting film and an imaged light pattern to produce an electrical signal from which can be derived the magnitude and phase of the two-dimensional Fourier transform of the image pattern. The sensor may also be used to produce other useful image functions.

The operation of the sensor can be explained with the aid of Figure 2. The sensor is fabricated on a  $\text{LiNbO}_3$  substrate which is a piezoelectric material. On this substrate is deposited a photoconducting film of CdS. An interdigital metal pattern is evaporated onto the CdS for the purpose of detecting and integrating the photocurrent. This pattern has the added function of sampling the image in one direction. Image sampling is required if base-band Fourier transform components are desired. The SAW are limited to high spatial frequencies. The spatial spectrum of the image must be translated into this band. To achieve a

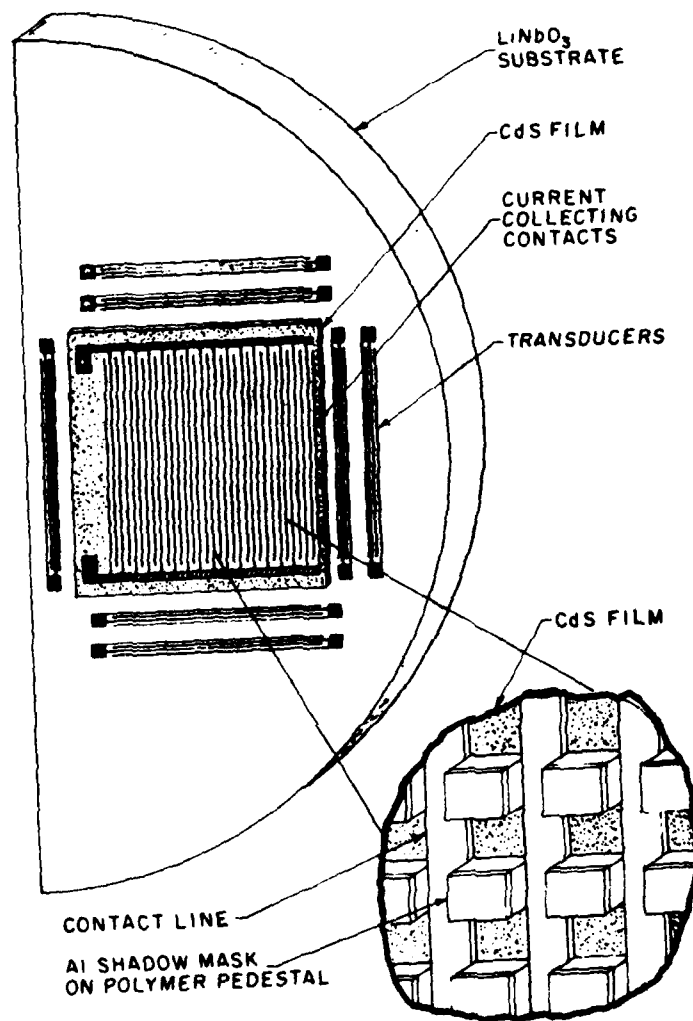


Figure 2 - Deft Sensor Construction

spatial sampling in the orthogonal direction the CdS can be overlaid with an Al shadow mask. As an alternative which has been used in more recent devices, only squares of CdS with the proper sample spacing are deposited. Interdigital transducers are used to generate two orthogonal surface acoustic waves in the substrate. The figure shows a number of transducers. However, only two are used to launch the orthogonal SAW's. The other transducers are redundant and not used.

To describe the operation of the sensor let  $g(t)$  and  $h(t)$  be the transducer input signals. Since  $\text{LiNbO}_3$  is piezoelectric an electrical field is produced across the sensor. In the  $x$  direction the field is proportional to  $g(t-x/v_x)$ . In the  $y$  direction the field is proportional to  $h(t-y/v_y)$ . The parameters  $v_x$  and  $v_y$  are the SAW velocities of propagation in the  $x$  and  $y$  directions. At the same time, an image is focused on the grid of CdS squares. A current is generated in each square proportional to the average light intensity over that square, and these currents are modulated by the product,  $g(t-x/v_x) h(t-y/v_y)$ , created by nonlinear mixing in the CdS film. The metal finger pattern sums each of the modulated current components over the grid. The resulting output current is  $o(t)$ . It is given by

$$o(t) = \iint \hat{I}(x,y) g(t-x/v_x) h(t-y/v_y) dx dy \quad (1)$$

The image  $\hat{I}(x,y)$  is the image focused on the sensor,  $I(x,y)$  modified by the grid sampling. The functions  $g(t)$ ,  $h(t)$  can be any arbitrary waveforms whose spectrum is within that of the interdigital transducers. This equation defines the function of the Deft sensor in its most general form.

A number of specific functions which can be realized by the sensor may have application in feature extraction. These will be discussed in the next section. A specific function which may have application in feature extraction is the two-dimensional Fourier transform. Since the performance of the Deft sensor in general applications can best be described in terms of its spectral response, computation of the two-dimensional Fourier transform using the sensor will be described here.

If  $g(t)$  and  $h(t)$  are sinusoids then  $o(t)$  yields components of the two-dimensional Fourier transform.

$$F(\omega_x, \omega_y) = \iint I(x,y) e^{-j(\omega_x x + \omega_y y)} dx dy$$

By varying the frequencies of the sinusoids different spatial frequencies can be probed. Let  $\rho_x, \rho_y$  be the grid spacing in the  $x$  and  $y$  direction respectively in meters/cycle. Then the temporal frequency which corresponds to the origin of the spatial transform is given by  $f_{o,x}, f_{o,y}$  where

$$f_{o,x} = v_x / \rho_x, \quad f_{o,y} = v_y / \rho_y \quad (2)$$

For example, if  $\rho_x = 0.1 \text{ mm/cycle}$ ,  $v_x = 3800 \text{ m/s}$  then  $f_{o,x} = 38 \text{ MHz}$ .

The number of spatial frequencies which can be probed by the sensor depends on the bandwidths of the two SAW transducers and the size of the CdS grid area. Let  $n_x$  and  $n_y$  be the number of CdS squares in the  $x$  and  $y$  direction respectively. Then for two spatial frequencies to be resolvable by the sensor, the total number of cycles of one frequency across the sensor must differ from the total number of cycles of the other by at least one cycle. For example, consider two  $x$ -directed spatial frequencies which differ by one cycle across the sensor. If  $n_x$  is the number of squares of CdS in the  $x$  direction then the difference in spatial frequency is  $1/(n_x \rho_x)$ . The difference in temporal frequency is then  $v_x/(n_x \rho_x)$ . Comparing this to equation (2) shows that the difference frequency,  $\Delta f_x$  is

$$\Delta f_x = f_{o,x} / n_x \quad (3)$$

Now let  $\nabla f_x$  be the temporal bandwidth of the  $x$  transducer. Then the number of resolvable frequencies in the  $x$ -direction is

$$r_x = \frac{\nabla f_x}{\Delta f_x} = \frac{\nabla f_x}{f_{o,x} / n_x} = n_x \frac{\nabla f_x}{f_{o,x}} \quad (4)$$

That is, the number of resolvable components in the  $x$  direction is  $n_x$  times the percentage bandwidth of the  $x$  transducer. For example, if  $\nabla f_x / f_{o,x} = 0.1$  and  $n_x = 200$  then there will be 20



resolvable Fourier components in the x direction. For  $\ell_x = 0.1$  mm/cycle this corresponds to a length of 20mm for the CdS portion of the sensor. In like manner, the number of resolvable frequencies in the y direction is given by

$$r_y = n_y \frac{v f_y}{f_{0,y}} \quad (5)$$

The number of resolvable Fourier components is an important parameter in applications such as feature extraction. It is used to characterize the operation of the Deft sensor even when a function other than the Fourier transform is being computed. For example, consider the more general function

$$u = \iint \hat{I}(x,y) g(x) h(y) dx dy \quad (6)$$

to be computed by the sensor. Because of the limits of integration  $g$  and  $h$  can be replaced by their periodic extensions  $[g]$  and  $[h]$ . These functions will be defined by an example. Assume that  $g(x) = x$  and that the  $x$  limits of integration are from  $-1$  to  $+1$ . Then the periodic extension of  $g(x)$  is shown in Figure 3.

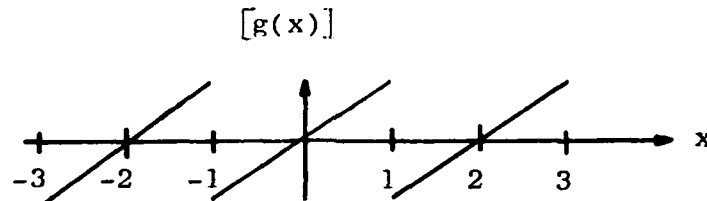


Figure 3 - Periodic Extension of  $g(x) = x$

The function  $[h(y)]$  is likewise defined with respect to the limits of integration in the  $y$  direction. Because  $[g]$  and  $[h]$  are periodic they may be expanded in complex Fourier series.

$$[g] = \sum_{k=-\infty}^{\infty} c_{g,k} e^{(jk\pi x/2)} \quad (7)$$

$$[h] = \sum_{k=-\infty}^{\infty} c_{h,k} e^{(jk\pi y/2)} \quad (8)$$

where  $j = \sqrt{-1}$ . Substituting these series in equation (6) and interchanging the order of integration and summation yields

$$u = \sum_{i=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} c_{g,i} c_{h,k} \iint \hat{I}(x,y) e^{j\pi(ix + ky)/2} dx dy \quad (9)$$

$$= \sum_{i=-\infty}^{\infty} \sum_{k=-\infty}^{\infty} c_{g,i} c_{h,k} F(i\pi/2, k\pi/2) \quad (10)$$

where  $F$  is the two-dimensional Fourier transform of  $\hat{I}$ . Since the Deft sensor transducer bandwidths are limited,  $u$  can be approximated by the truncated series

$$\hat{u} = \sum_{i=-n_x}^{n_x} \sum_{k=-n_y}^{n_y} c_{g,i} c_{h,k} F(i\pi/2, k\pi/2) \quad (11)$$

where  $n_x+1$ ,  $n_y+1$  are the number of resolvable Fourier components along the  $\omega_x$  and  $\omega_y$  axis respectively. The error in computing  $u$  with the sensor is

$$u - \hat{u} = \sum_{|i| > n_x} \sum_{|k| > n_y} c_{g,i} c_{h,k} F(i\pi/2, k\pi/2) \quad (12)$$

This error depends on  $n_x$ ,  $n_y$  and on the high frequency content of the image  $\hat{I}$  and of the kernel  $g(x)h(y)$ . Hence, given a sensor with parameters  $n_x$ ,  $n_y$  it is possible to determine which kinds of images may be viewed and which kinds of kernels may be utilized.

The Deft sensor is further described in a number of references <sup>2,3,4</sup>.

### C. Elastobirefringent Light Valve

A two-dimensional acoustic processor utilizing an elastobirefringent light valve has been constructed and experiments conducted <sup>5,6</sup>. The experimental setup is shown in Figure 4. The light valve consists of a fused quartz cell. On the edges of this cell are attached two orthogonal transducers. When

these transducers are excited strain waves travel through the bulk of the cell. The total strain is the linear superposition of the two traveling strain waves. An image is focused on the region of the quartz where the two waves are superimposed.

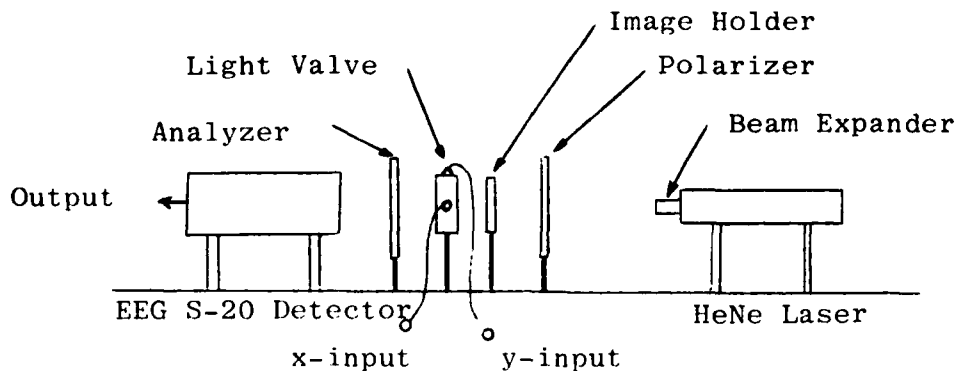


Figure 4 - Elastobirefringent Light Valve: Experimental Set-Up

Due to the strain induced birefringence, each ray of light splits into two orthogonally polarized phase velocities. It can be shown that the intensity modulation of each polarized wave has a component proportional to the strain squared. The image is produced by illuminating a transparency containing the image information with a well-collimated beam which has been polarized. The output of the light valve will have a desired component polarized orthogonal to the orientation of the input polarizer. This component is passed by the analyzer which is just another polarizer. The undesirable direct component is blocked by the analyzer. The desired component is then detected by a photodetector which performs an integration over the entire image. A chromium sampling grid is placed on the quartz cell to sample the image and translate its spatial spectrum into the passband of the transducers.

The function which can be computed by this processor can be described by equation (1) which is also used to describe the

Deft sensor. Hence, functionally this processor is equivalent to the Deft sensor. The differences between the two are primarily in implementation and in present and projected future performance. Concerning implementation, the light valve is a larger physical structure than the Deft sensor. The image source must be a transparency. By contrast, the Deft sensor can view either a transparency, a photograph or a real-world image since it operates much like a conventional camera. The elastobirefringent light valves which have been built to date exhibit a strong standing wave pattern which is a function of transducer frequency. This effect has limited their usefulness. Potential solutions to this problem are presented in Section V.

This light valve does have a couple of potential advantages compared with the Deft sensor. In the bulk mode devices it is not necessary to physically attach a metal pick-up grid. In the Deft sensor this grid has an undesirable effect of damping the SAW. This limits the practical size of Deft sensors since larger sensors would suffer a large, undesirable SAW attenuation across the sensor surface. Hence, the bulk devices could potentially be larger resulting in a greater number of resolvable Fourier components. A second advantage is that electrical feed-through is not a problem with the bulk devices. With the Deft sensor feed-through from the input transducer drive to the integrating metal grid is a problem.

However, the bulk device has a number of disadvantages when compared with the Deft sensor. Since bulk mode operation is required, the image must be focused with sufficient depth-of-focus so that the image is in focus throughout the 1-2mm thickness of the quartz cell. Higher acoustic fields are required to operate the bulk device. The light valves built to date use glued-on transducers. This is not desirable, particularly at higher frequencies where the transducers are smaller and more brittle. Finally, the physical non-linearity which is utilized in the bulk light valve is a much smaller effect than

that used in the Deft sensor. There is a large, undesirable unmodulated signal which feeds through the light valve. The purpose of the analyzer is to block this signal. However, it is difficult to fabricate sheet polarizers which are uniform enough over the 1 cm. illuminated length of the quartz cell. Hence, some of this feed-through signal will pass the analyzer and could saturate the detector. (This feed-through signal is analogous to the electrical feed-through encountered in the Deft sensor.)

#### D. Thomson - CSF Sensor

The Thomson - CSF sensor is a SAW device and has features in common with the Deft sensor. The device can be described as a two-dimensional separated media semiconductor convolver. It employs two pairs of SAW transducers deposited on a  $\text{LiNbO}_3$  substrate. An  $8 \times 8 \text{ mm}^2$  matrix of p-n vidicon type diodes is pressed on 3000 Å high randomly distributed posts. Output signals are picked up between the back electrode of the semiconductor and a semi-transparent ground electrode deposited on the bottom face of the piezoelectric medium. The device is shown in Figure 5. The x and y center frequencies are  $f_{o,x}$  and  $f_{o,y}$ . The useful output frequency then appears at frequency  $2(f_{o,x} + f_{o,y})$ .

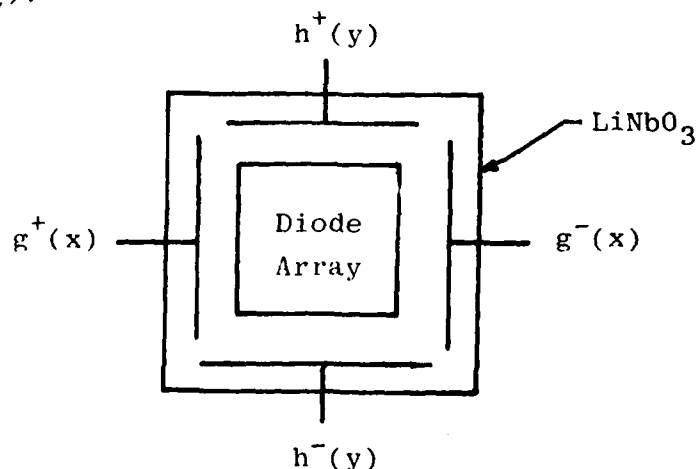


Figure 5 - Thomson - CSF Sensor

To generate the two-dimensional Fourier transform of an image focused on the diode array, the x input signals are chosen to be FM ramps of opposite slopes  $\omega_x^\pm = \omega_x \pm ut$ . The y input signals are chosen to be CW signals with frequencies  $\omega_y^\pm = \omega_y \pm 1/2 \Delta \omega_y$ . (That is, one x orientated transducer gets  $\omega_x^+$  and the other x orientated transducer gets  $\omega_x^-$ . The same is true for y orientated transducers.) The instantaneous frequencies of these waveforms are shown in Figure 6 (solid lines).

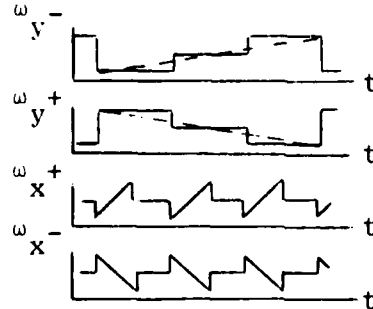


Figure 6 - Transducer Instantaneous Frequencies

With these inputs, the output signal component at  $2(\omega_x + \omega_y)$  can be shown to have a modulation of the form

$$o(t) = \iint \hat{I}(x, y) e^{j(-2utx/v_x + \Delta \omega_y/v_y)} dx dy \quad (13)$$

where  $v_x$ ,  $v_y$  are the SAW velocities of propagation. During a sweep of the x, FM ramps an entire row of the transform is read from the sensor. This sweep time need be no more than twice the propagation time across the sensor. Hence, data can be outputted potentially much faster than can be achieved using the Deft sensor. In the Deft sensor the x and y input waves must be CW signals which must propagate across the sensor before a single transform component is outputted. The frequencies of these waves are then stepped to output another component, etc. The Thomson - CSF sensor approach would be most useful if a raster-scan of the transform is desired. This is shown in Figure 7 (solid lines).

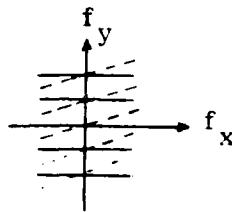


Figure 7 - Raster Scan of Transform Plane

If instead of stepping the  $y$  frequencies, the  $y$  frequencies are also ramps (dashed lines, Figure 6) then the transformed plane can also be raster-scanned along oblique lines with variable tilt angle (Figure 7, dashed lines). For example, it would be possible to scan the transform along radial lines passing through the transform origin. This is shown in Figure 8.

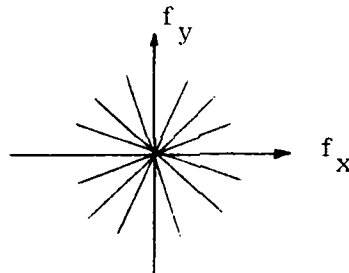


Figure 8 - Radial-Scanned Transform

This method of scanning will be shown in Section III to have application in feature extraction.

The primary difference between the Deft sensor and the Thomson - CSF sensor is the use of four vs. two transducers and the choice of waveforms leading to a fast scan of the transform. If random samples or samples at only a few spatial frequencies are all that is required then the Deft sensor would function just as rapidly as the Thomson - CSF sensor. It would be possible to operate a Deft sensor in the same manner as the Thomson - CSF sensor by also using four transducers and inputting FM ramp waveforms. Speed performance would then

equal the Thomson - CSF device. This modification has not been attempted to date. Note that the method used to speed-up data output is not applicable to the computation of the more general operator of equation (1).

The Thomson - CSF sensor is described in reference 7.



### III. FEATURE EXTRACTION TECHNIQUES FOR ACOUSTO-OPTIC TECHNOLOGY

#### A. Introduction

During the execution of this contract a survey of techniques for feature extraction was carried out. The purpose of this survey was to determine techniques suitable for implementation using acousto-optic devices. The results of this survey are detailed in this chapter. From the techniques surveyed, two methods appear to offer promise both from an algorithmic standpoint and also because of suitability of implementation. These methods will be termed the method of invariant Fourier signatures (IFS) and the method of invariant moment signatures (IMS). Because of their promise, additional algorithmic development was carried out during the course of this contract. This work is reported in this section.

This section is divided into two parts. The first part is a general discussion of feature extraction. The second part is a comparison of some methods for feature extraction.

#### B. General Concepts

A general model for a feature extraction system is presented in Section 1B. This model is shown in Figure 1. Briefly, the transducer converts the image input light pattern to electrical signals. The preprocessor reduces the information content of the image to a more manageable size. The output of the preprocessor is then passed to the decision processor where this information is used to classify the photo image.

In Section II the operation and function of three A-0 devices were considered. The general function performed by these devices is given by equation (1). This equation implies that the two-dimensional image information is processed by the sensor in such a way to result in a one-dimensional signal of reduced complexity. Hence, these devices could be potentially used to implement the combined function of transducer and preprocessor. Therefore, it is important to consider specific

examples of equation (1) and determine their utility in pre-processing. This will be done in this section.

The A-0 devices considered do not seem useful for implementing the decision processor. There are a number of reasons for this. First of all, the output of the preprocessor will be an electrical signal. The A-0 devices require an image input. Hence, the data format is not compatible. A survey of the types of algorithms commonly used for classification reveals that most of these cannot be implemented efficiently using A-0 devices. In addition, they require greater accuracy than can be achieved with A-0 devices. But most importantly, it is not necessary to use A-0 devices for classification since the image information content has been reduced by the preprocessor to the point where a modest digital processor can handle this function. A survey of conventional classification methods is given in reference 8. An application of classification in aircraft identification is given in reference 9.

Before developing specific preprocessor functions some desirable properties of the transducer plus preprocessor will be discussed.

The first desirable property is feature isolation. Consider a typical aerial image of natural terrain in which may be located one or more man-made features to be detected. The natural terrain can be considered to be "noise" while the feature is a "signal". One of the functions of the preprocessor is to filter this signal from the noise. Now in a typical aerial photograph, most of the photo will be noise. If the transducer were to view the entire photograph then the signal-to-noise ratio would be small and the preprocessor function would be more difficult to implement. A simple but effective way to improve the signal-to-noise ratio would be to limit the field of view of the transducer to only a portion of the photo and then scan the photo to search for features. Limiting the field of view has the effect of reducing the noise without reducing the signal as long as the feature to be detected remains completely within the field of view.

By scanning the photo is meant that the transducer views a portion of the photo and that portion is preprocessed. The transducer then views another portion of the photo and preprocessing is repeated for that portion. This sequence of steps is continued until all portions of the image have been viewed. Scanning may or may not involve overlapping views. Scanning a square photo with a square aperture and no overlapping is shown in Figure 9. The same scan but with 50% overlap in both directions is shown in Figure 10. To simplify presentation the views for the overlap case are shown in four parts. These parts would overlap each other. The boundary of the photo is shown dashed. Now consider a feature which is completely enclosed by the circle shown in Figure 9. (The angular orientation of the feature is arbitrary.) Then if the photo was scanned as in Figure 9 the feature would lie across the boundary of four views and its detection would be difficult. However, if 50% overlapping of views is used as in Figure 10 then the feature lies totally within one view (in this case, view 48). As long as the radius of the circle was less than or equal to one quarter of the length of the square comprising each view and overlap was 50% the circle would lie completely within one view. Note that (neglecting missing squares at the edges of the photo) the number of views has increased by a factor of four.

These comments can be generalized. Let  $r$  be the radius of the smallest circle which completely encloses the feature. Obviously,  $r$  is independent of the angular orientation of the feature in the plane of the image. Assume that the aperture of the transducer imaged on the photo is square with length  $\ell$ . Figure 11 shows one such aperture (large square). In the middle of this square is a smaller square. The area enclosed by this square is the locus of points of all circles of radius  $r$  which lie completely within the large square. Each view has such a small square associated with it. In order that the

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16

Figure 9 - Image Scanning: No Overlap

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	16

17	18	19	20
21	22	23	24
25	26	27	28

29	30	31
32	33	34
35	36	37
38	39	40

41	42	43
44	45	46
47	48	49

Figure 10 - Image Scanning: 50% Overlap

feature lie totally within a view regardless of the position of the feature within the image, the views must overlap in such a way that the entire area of the photo is covered by the union of all these small squares. In that case, the total number of views must equal at least

$$\text{total number of views} = \ell_T^2 / (\ell - 2r)^2 \quad (14)$$

where  $\ell_T$  is the length of the square photo. That is, the total number of views is greater or equal to the total area of the photo divided by the area of a square of length  $\ell - 2r$ .

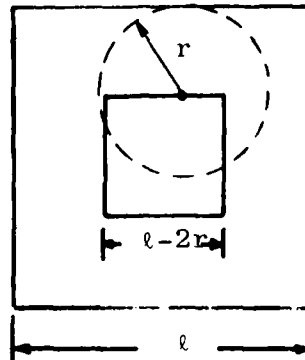


Figure 11 - Overlap Geometry

The signal-to-noise ratio (SNR) of the view will be defined to equal the area of the feature divided by the area of the view, Hence,

$$\text{SNR} \leq 2\pi r^2 / \ell^2 \quad (15)$$

with maximum occurring when  $\ell = 2r$ . However, in that case the total number of views required is infinite. Hence, there must be a tradeoff between SNR and total number of views to be processed. If the ratio  $r/\ell_T$  is very small (small features) then a large number of views are required to achieve good SNR. For example, assume that  $2r = \ell_T/50$  and that  $\ell = 12r$  then  $\text{SNR} \leq \pi/72$  and the total number of views is 100.

In practice, the total number of views can be reduced somewhat by allowing some of the feature to be masked by the

transducer aperture. That is,  $r$  is chosen to be somewhat smaller than the value implied from the dimensions of the feature. To develop a general formula let  $\ell = k_1 r$ . Now reduce  $r$  to  $k_2 r$ . Then the total number of views is reduced by a factor of  $(k_1 - 1)^2 / (k_1 - k_2)^2$ . In the previous example  $k_1 = 12$ . Let  $k_2 = 0.8$ . That is, overlap will be reduced so that, in the worst case, only 80% of the largest feature dimension will be within a view. In this case the total number of views required is reduced by a factor of 0.9646. For a fixed value of  $k_2$  this reduction will be greater for smaller values of  $k_1 > 2$ . For example, if  $k_1 = 4$  and  $k_2 = 0.8$  then the total number of views is reduced by a factor of 0.8789.

In terms of implementation, a single transducer and prefilter could be used to scan the photo in a number of steps. If high-speed operation was required more than one transducer and prefilter could view separate views of the photo concurrently.

Another desirable property of the prefilter will now be discussed. By definition, the prefilter takes the large amount of information of the photo encoded as pixels and reduces it to a much smaller set of numbers. This process can be termed "dimensionality reduction". Ideally, the prefilter output should have the following properties:

- 1) Members of each feature class should show less variability than was the case before the prefilter.
- 2) The relative separation of each feature class should be increased.
- 3) There should be dimensionality reduction.

Consider first property 1). As an example of a feature class let the class include all square buildings in the photo. In this case, a square building is in the feature class regardless of its angular orientation or scale (size). It would be desirable for the prefilter to have the property that if the photo contains a square building in any position, orientation or scale then the prefilter output will contain a signal or

signature which is invariant to position, orientation or scale. Such a prefilter will be said to implement a position, rotation and scale invariant transformation. In certain cases it may be desirable that the transformation is only position and rotation invariant. (Example: small buildings are to be discriminated from large buildings.)

The best-known invariant transformation is the autocorrelation function. The first-order autocorrelation function defined in one dimension is

$$\phi_g^1(\Delta) = \int g(x)g(x+\Delta) dx \quad (16)$$

this function is invariant to translation.

Consider now property 2). As an example, assume that the prefilter input is a photo containing a square building. This building may be in any position, orientation and scale. The set of all outputs constitutes a class at the prefilter output. In this idealized situation, if the prefilter is invariant to translation, rotation and scale then this class contains but one output or signature. A more realistic consideration would have to include such effects as variation of building materials and color and lighting. Another class could be determined by considering all outputs for all possible orientations of "round storage tanks" in a photo. Now if the differences in all signatures of one class from all signatures in the other class is increased by prefiltering then relative separation has increased. This might be the case, for example, if noise, such as natural features, is filtered out by the prefilter.

However, in some cases relative separation may decrease. For example, consider the autocorrelation function. It is well-known that

$$F\{\phi\} = |F\{g\}|^2 \quad (17)$$

where  $F\{\}$  is the Fourier transform. Hence, two functions whose Fourier transforms have the same magnitude cannot be separated by using the autocorrelation functions. Functions which cannot

be separated include not only those which differ by a linear phase term (i.e., the translated versions of the same function) but also those which differ by a nonlinear phase term.

A function which avoids this difficulty is the higher-order autocorrelation function

$$\phi_g^n(\Delta_1, \Delta_2, \dots, \Delta_n) = \int g(x)g(x+\Delta_1)g(x+\Delta_2)\dots g(x+\Delta_n) dx \quad (18)$$

which can be shown to be unique for  $n \geq 2$  except for translation<sup>10</sup>. However, note that the dimensionality has been increased. This conflicts with the third desirable property that there should be a dimensionality reduction.

In general, it is not possible to satisfy all three desirable features of a prefilter at the same time. It would seem to be important that there be a dimensionality reduction since, otherwise, the amount of information presented to the decision processor would be enormous.

Later in this chapter two classes of algorithms are presented which satisfy properties 1) and 3) but violate property 2). This would seem to be the lesser evil since the decision processor receives less information which has been formatted in such a way that the decision process is simplified. Since property 2) is violated, a square building, for example, may be mistaken for a round storage tank. However, this is less important than not detecting the feature. It is always possible to refine the decision process by human intervention. A feature extractor which signals the presence of a feature and indicates with some certainty what that feature is could be very useful for automatic aerial photo screening.

### C. Transform Methods

Because of the scope of this project, this section will concentrate on prefilters which utilize transforms which can be implemented as a special case of equation (1). Transforms which will be considered are the two-dimensional Fourier transform, the two-dimensional image moments and the two-



dimensional Hadamard transform. Applications of the Fourier and Hadamard transform to image processing is surveyed by Pickholtz<sup>11</sup>. The application of moments to image processing is discussed in a number of papers<sup>9,12,13</sup>.

#### 1. Hadamard Transform

The Hadamard transform of an image which has been sampled and represented by the  $n \times n$  matrix of pixel values  $I(x_i, y_j)$  is

$$F(i, j) = \frac{1}{n} H_n I H_n \quad (19)$$

the  $n \times n$  matrix  $H_n$  consists of elements which are either +1 or -1. The Hadamard transform is orthonormal. As a result, the image is decomposed into basis images which are the two-dimensional Walsh functions. The rows and columns of the matrix  $H_n$  (which is symmetric) consists of the one-dimensional Walsh functions of order  $n$ . For  $n = 8$  these Walsh functions are shown in Figure 12.

		<u>Sequency</u>
wal (0,t)	+++++++	0
wal (1,t)	++++----	1
wal (2,t)	++-----++	1
wal (3,t)	++--++--	2
wal (4,t)	+---+---+	2
wal (5,t)	+---+---+	3
wal (6,t)	+--+--+--	3
wal (7,t)	+--+--+--	4

Figure 12 - First Eight Walsh Functions

The notation + means +1 while - means -1. These functions are defined on the interval  $0 \leq t \leq 1$  and may be periodically extended to span the real line. The Walsh functions may be thought of as square waves. Wal (6,t) is shown in Figure 13 a). For each of these functions, every occurrence of a transition from + to - or from - to + is called a zero crossing. One half the number of zero crossings of these functions is termed the sequency of the function. Sequency is analogous to frequency of sine and cosine functions.

The transform  $F(i,j)$  is a decomposition of the image into its orthonormal Walsh basis images. The potential advantage of this representation is that it may be possible to accurately approximate the image by using only a limited number of its Hadamard transform components. If so, this would facilitate feature extraction since the output of the Hadamard prefilter would be of lower dimensionality than the input.

The Hadamard transform also appears useful because of its ease of implementation. Because  $H_n$  only consists of +1 or -1 entries the matrix product represented by equation (19) can be computed without any multiplies. This is particularly useful in a digital processor where multiplications require more computation than do adds and subtracts. A high-speed digital processor which computes the Hadamard transform of an image for feature extraction is described in reference 14.

One potential disadvantage of the Hadamard transform is that it is not invariant to translation, rotation or scale of a feature in the image.

Consider now the computation of the Hadamard transform using acousto-optic devices. The potential application of the Deft sensor to computing the Hadamard transform has previously been discussed<sup>3</sup>. Consider the general function performed by A-O devices as defined by equation (1). Because the image  $I(x,y)$  has been sampled by the sampling grid to produce  $\hat{I}(x,y)$ , the functions  $g$  and  $h$  must be shifted in frequency into the band occupied by the image. Hence,  $g$  and  $h$  cannot be base-band Walsh functions. The appropriate choice is

$$g(t) = \sin \omega_{o,x} t \text{ wal}(i,t) \quad (20)$$

$$h(t) = \sin \omega_{o,y} t \text{ wal}(j,t) \quad (21)$$

where  $\omega_{o,x}$  and  $\omega_{o,y}$  are the frequencies which will produce SAW's with wavelengths equal to the metal grid pattern in the  $x$  and  $y$  directions, respectively. These waveforms are shown in Figure 13.

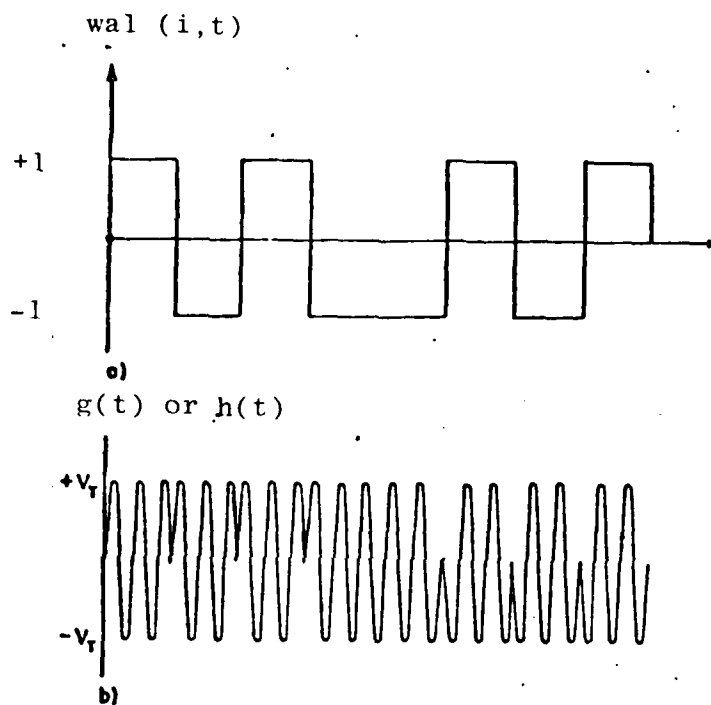


Figure 13 - Modulation of Sinusoid by a Walsh Function  
a) Walsh function, b) Modulated Sinusoid

However, to be realistic, it is important to consider the limitations imposed by the bandwidth of the A-0 device transducers. Since the Walsh functions are square waves they require unlimited bandwidth for exact representation. However, most of the spectral energy of these waveforms will lie below some maximum frequency. As long as the transducers pass frequencies within this bandwidth the A-0 device can function as a Hadamard transformer. Walsh function spectrums have been derived in the form of recursive formulas<sup>15</sup>. In order to present these formulas some additional notation is required. First, define the cal and sal functions

$$\text{cal}(n/2, t) = \text{wal}(n, t); n \text{ even.} \quad (22)$$

$$\text{sal}((n+1)/2, t) = \text{wal}(n, t); n \text{ odd.} \quad (23)$$

Now define the Fourier transforms of these functions to be

$$C(n, f) = \int_{-1/2}^{+1/2} \text{cal}(n, t) e^{-j\omega t} dt \quad (24)$$

$$S(n, f) = \int_{-1/2}^{+1/2} \text{sal}(n, t) e^{-j\omega t} dt \quad (25)$$

The interval  $-1/2 \leq t \leq 1/2$  can be thought of as the time that the Walsh function acoustic wave will be on the active sensor area.

With these definitions the recursive formulas for  $C(n, f)$  and  $S(n, f)$  are given by

$$C(n, f) = \begin{cases} (-1)^{[n/1]} \cos(\pi f/2) C(n/2, f/2); n \text{ even} \\ j(-1)^{[n/2]} \sin(\pi f/2) S((n+1)/2, f/2); n \text{ odd} \end{cases} \quad (26)$$

$$S(n, f) = \begin{cases} -(-1)^{[n-1]/2} \cos(\pi f/2) S(n/2, f/2); n \text{ even} \\ -j(-1)^{[n-1]/2} \sin(\pi f/2) C((n-1)/2, f/2); n \text{ odd} \end{cases} \quad (27)$$

where  $[n/2]$  means the largest integer equal to or smaller than  $n/2$ .

Using these formulas the Fourier transform of the Walsh functions can be easily computed. The first few of these are shown in Tables 1 and 2. Each of these functions is a product of sines and cosines multiplied by a term of the form  $\sin^2(\pi f/k)/(\pi f/k)$ . The bandwidth of the Walsh functions is

<u>n</u>	<u>j S(n, f)</u>
0	-
1	$\frac{\sin^2(\pi f/2)}{\pi f/2}$
2	$\frac{-\cos(\pi f/2) \sin^2(\pi f/4)}{\pi f/4}$
3	$\frac{-\sin(\pi f/2) \sin(\pi f/4) \sin^2(\pi f/8)}{\pi f/8}$
4	$\frac{-\cos(\pi f/2) \cos(\pi f/4) \sin^2(\pi f/8)}{\pi f/8}$
5	$\frac{-\sin(\pi f/2) \cos(\pi f/4) \sin(\pi f/8) \sin^2(\pi f/16)}{\pi f/16}$
6	$\frac{\cos(\pi f/2) \sin(\pi f/4) \sin(\pi f/8) \sin^2(\pi f/16)}{\pi f/16}$
7	$\frac{\sin(\pi f/2) \sin(\pi f/4) \cos(\pi f/8) \sin^2(\pi f/16)}{\pi f/16}$
8	$\frac{-\cos(\pi f/2) \cos(\pi f/4) \cos(\pi f/8) \sin^2(\pi f/16)}{\pi f/16}$

Table 1 - S(n, f) vs. n

<u>n</u>	<u>C(n, f)</u>
0	$\frac{\sin(\pi f)}{\pi f}$
1	$\frac{\sin(\pi f/2) \sin^2(\pi f/4)}{\pi f/4}$
2	$\frac{-\cos(\pi f/2) \sin(\pi f/4) \sin^2(\pi f/8)}{\pi f/8}$
3	$\frac{\sin(\pi f/2) \cos(\pi f/4) \sin^2(\pi f/8)}{\pi f/8}$
4	$\frac{-\cos(\pi f/2) \cos(\pi f/4) \sin(\pi f/8) \sin^2(\pi f/16)}{\pi f/16}$
5	$\frac{-\sin(\pi f/2) \sin(\pi f/4) \sin(\pi f/8) \sin^2(\pi f/16)}{\pi f/16}$
6	$\frac{-\cos(\pi f/2) \sin(\pi f/2) \cos(\pi f/8) \sin^2(\pi f/16)}{\pi f/16}$
7	$\frac{-\sin(\pi f/2) \cos(\pi f/4) \cos(\pi f/8) \sin^2(\pi f/16)}{\pi f/16}$
8	$\frac{-\cos(\pi f/2) \cos(\pi f/4) \cos(\pi f/8) \sin(\pi f/16) \sin^2(\pi f/32)}{\pi f/32}$

Table 2 - C(n, f) vs. n

determined by this term. In general, sal transforms from  $S(2^{n+1}, f)$  through  $S(2^{n+1}, f)$  contain a term of the form  $\sin^2(\pi f/2^{n+2})/(\pi f/2^{n+2})$  while cal transforms from  $C(2^n, f)$  through  $C(2^{n+1}-1, f)$  contain a term of the form  $\sin^2(\pi f/2^{n+2})/(\pi f/2^{n+2})$ . The half power point of this envelope function occurs at

$$f_{1/2} = .5816 \times 2^{n+2} \quad (28)$$

(that is  $\sin^2(\pi f_{1/2}/2^{n+2})/(\pi f_{1/2}/2^{n+2}) = 1/\sqrt{2}$ .) If the number of Hadamard transform components along one axis is doubled then the corresponding transducer bandwidth must also be doubled. An identical statement can be made concerning bandwidth requirements for an A-0 Fourier transformer.

Consider now the bandwidth required to excite Walsh functions on the A-0 sensor. The sequency of  $S(2^{n+1}, f)$  and  $C(2^{n+1}, f)$  is  $2^{n+1}$ . A sine or cosine wave defined on the same interval (i.e.,  $0 \leq x \leq 1$ ) also has a sequency (i.e., number of zero crossings). For example, a sine or cosine wave with sequency  $2^{n+1}$  has frequency

$$f_s = .5 \times 2^{n+2} \quad (29)$$

From the discussion above, the frequency of this sinusoid is slightly lower than the half power point of  $S(2^{n+1}, f)$  and  $C(2^{n+1}, f)$ . The maximum of the envelope occurs at  $f = .3711 \times 2^{n+2}$  while the first zero of this function which is not at the origin occurs at  $f = .6366 \times 2^{n+2}$ . This is shown in Figure 14.

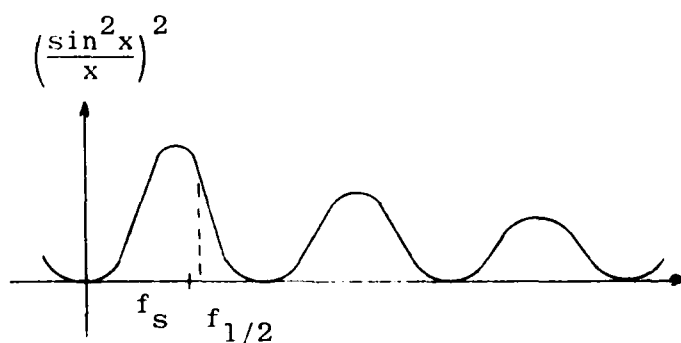


Figure 14 - Walsh/Fourier Bandwidth Relationship

However, because of the energy contained in the sidelobes of the envelope function it would seem advisable to require a bandwidth at least 2 or 3 times that required for sinusoids of the same frequency.

Because of the requirement of extra bandwidth, the use of A-0 devices to compute Hadamard transforms is not as attractive as using these same devices to compute Fourier transforms. The utility of the Hadamard transform becomes important in implementations where it is costly to include multiplications. However, the A-0 devices discussed in Section II are ideally suited for functional multiplication. Hence, using these devices to compute Hadamard transforms is probably not their best application.

## 2. Matched Filtering

The matched filter has long been used in communications systems as a means to filter a signal corrupted by additive noise. The extension of the matched filter to a two-dimensional signal (image) was first proposed by Vander Lugt<sup>16,17</sup>. Briefly, consider an image  $I(x,y)$  consisting of signal  $s$  plus noise  $n$ .

$$I(x,y) = s(x,y) + n(x,y) \quad (30)$$

The matched filter is a linear filter with impulse response  $h(x,y)$  which filters the image in such a way to maximize the signal-to-noise ratio at the output. Let  $F(\omega_x, \omega_y)$ ,  $S(\omega_x, \omega_y)$ ,  $N(\omega_x, \omega_y)$  and  $H(\omega_x, \omega_y)$  be the two-dimensional Fourier transforms of  $I$ ,  $s$ ,  $n$  and  $h$  respectively. The noise is usually characterized by its power spectrum  $NN^*$ . The signal-to-noise ratio is defined to be

$$\text{signal-to-noise} = \frac{\left| \iint S H e^{j(\omega_x x + \omega_y y)} d\omega_x d\omega_y \right|^2}{\iint NN^* H H^* d\omega_x d\omega_y} \quad (31)$$

The signal-to-noise ratio is maximized over the filter output space if

$$H(\omega_x, \omega_y) = \frac{S^*(\omega_x, \omega_y)}{NN^*} \quad (32)$$



the filter output is

$$m(x,y) = \iint F H e^{j(\omega_x x + \omega_y y)} d\omega_x d\omega_y \quad (33)$$

If the noise is white, ( $NN^*$ , a constant), then the matched filter output is simply the correlation between the image and the desired object to be detected.

Now if the image contains the signal but offset to position  $\Delta x, \Delta y$  then the correlation in the output will be shifted to  $-\Delta x, -\Delta y$ . There will be a peak in the output at this position. If the image contains multiple signals at various positions within the image then the output will also contain a number of correlation peaks. The matched filter is translation invariant in a limited sense. If the image contains a signal (feature) then the output will contain a correlation peak. However, if the signal is translated then the output peak is also translated. Strictly speaking, the output would be different (i.e., translated) so that the matched filter is not translation invariant. However, it is invariant in the sense that the correlation peak will occur somewhere only if the image contains the signal. Another way of looking at it is to say that the matched filter preserves the positional information about the signal. There is no dimensionality reduction since the matched filter output is a function of  $x$  and  $y$ . However, signal detection is improved.

Referring back to Section III A, if the image is scanned and broken into a number of views it may not be required to locate the position of a feature within a view. Rather,, the detection of a feature somewhere in the view may be adequate. Hence, the matched filter may preserve too much information. Perhaps by discarding this information a simpler implementation may result.

Now consider implementation of the matched filter using A-0 devices. When the two-dimensional matched filter was first discussed by Vander Lugt he proposed implementing it using coherent optics. In this implementation the image is

first transformed using a Fourier transformer lens. The matched filter is then implemented in the Fourier domain by a spatial filter. The filtered transform is then passed through a second Fourier transformer lens. The output image contains bright spots corresponding to signals in the input image. This is a rather natural implementation for an optical processor since the various lenses and filters can be lined up on an optical bench and the information passes through the system in a parallel fashion. That is, the data stream is a two-dimensional light pattern throughout the processor.

Because A-0 devices can compute two-dimensional Fourier transforms, an analogous system could be built using these devices. A block diagram of such a system is shown in Figure 15. The image is viewed by an A-0 device such as a Deft sensor. The sensor is used to compute the two-dimensional Fourier transform of the image. The transform is raster scanned by properly addressing A-0 sensor No. 1. This sensor will have two outputs which are the real and the imaginary part of the Fourier transform. These transform components are multiplied by the corresponding components of the matched filter which are stored in a memory. A complex multiplication operation consisting of four real multiplies is required. The weighted transform components are then input to two CRT's. Since the transforms are raster scanned, the data stream is in the correct format for the CRT. The real and imaginary parts of the Fourier transform, weighted by the matched filter, are displayed on CRT No. 1 and No. 2 respectively. These functions are then viewed by A-0 devices No. 2 and No. 3. It may be possible to directly attach these sensors to fiber optic faceplate CRT's. Since the functions on the CRT's must remain constant for the period of time required to scan out the inverse transforms from the A-0 devices, storage CRT's are required. The output of A-0 devices Nos. 2 and 3 are then A/O converted and input to the decision processor. The decision processor searches for correlation peaks in the

matched filter output and compares these peaks with a threshold valve in order to determine if the feature is contained in the image.

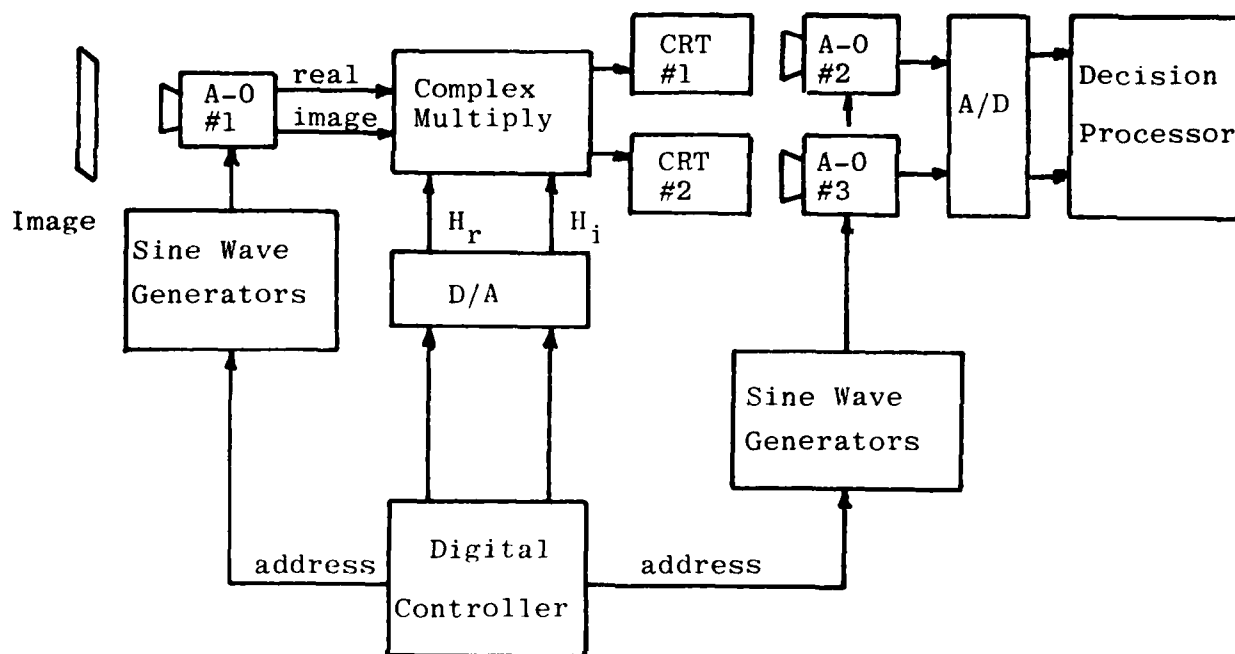


Figure 15 - Matched Filter: A-0 Implementation

A disadvantage of this implementation is the requirement of three A-0 devices and the need to generate a pictorial representation of the weighted image transform in order to utilize the second two A-0 devices.

In order to avoid this complexity, consider the alternative of computing the matched filter in the image domain rather than the transform domain. The matched filter output is the convolution of  $I(x,y)$  with  $h(x,y)$ . That is,

$$m(x,y) = \iint I(\alpha,\beta)h(\alpha-x,\beta-y)d\alpha d\beta \quad (34)$$

In order to compute this function by use of an A-0 device which implements equation (1) the function  $h(x,y)$  must be decomposable into

$$h(x,y) = g(x)h(y) \quad (35)$$

However, this will not generally be the case for arbitrary features to be detected. Hence, this approach cannot be taken.

In order to avoid the requirement for an inverse transform a third approach can be taken. As has been pointed out, the matched filter output preserves positional information which may not be required. Perhaps some implementation advantage can be gained by dispensing with this information. To this end, consider the matched filter for the autocorrelation function of  $I(x,y)$ . The transform of this function is given by

$$FF^* = (S+N)(S+N)^* = SS^* + SN^* + S^*N + NN^* \quad (36)$$

If the signal and the noise can be considered uncorrelated then  $SN^* = S^*N = 0$  and

$$FF^* = SS^* + NN^* \quad (37)$$

In this case the signal is the autocorrelation function and the noise is  $NN^*$ . The function  $FF^*$  consists of signal plus added noise. Hence, the matched filter can be used to filter the signal from the noise. In this case the matched filter is given by

$$\hat{H}(\omega_x, \omega_y) = \frac{(SS^*)^*}{(NN^*)(NN^*)^*} = \frac{(SS^*)^*}{(NN^*)^2} \quad (38)$$

The filter output is

$$\hat{m}(x,y) = \iint FF^* \hat{H} e^{j(\omega_x x + \omega_y y)} d\omega_x d\omega_y \quad (39)$$

However since  $F$  has been replaced by  $FF^*$ , any correlation peak indicating the signal or feature with transform  $SS^*$  will occur at  $x = y = 0$  at the matched filter output. Hence, it is only necessary to compute

$$\hat{m}(0,0) = \iint FF^* \hat{H} d\omega_x d\omega_y \quad (40)$$

This function is simpler to implement using A-0 devices. A functional block diagram is shown in Figure 16. The image is viewed by an A-0 device. The device again is used to compute the two-dimensional Fourier transform of the image. The sensor electronics are designed to output the magnitude of the Fourier transform rather than the real and imaginary parts. Computing the magnitude actually required less electronics than computing the real and imaginary part. This is because, in the second case, phase information must be preserved and a synchronous detector is required. Such a detector for a Deft sensor is described in reference 18, Section II. In the matched filter application the A-0 device is scanned over the range of spatial frequencies of interest. The sensor output is squared and multiplied by the corresponding values of the matched filter  $\hat{H}$ . Both the sensor output and  $\hat{H}$  values are real which simplifies this computation. The output of the second multiplier is then integrated over all spatial frequencies. The integrator output is compared with a threshold value to determine if the signal (feature) is present.

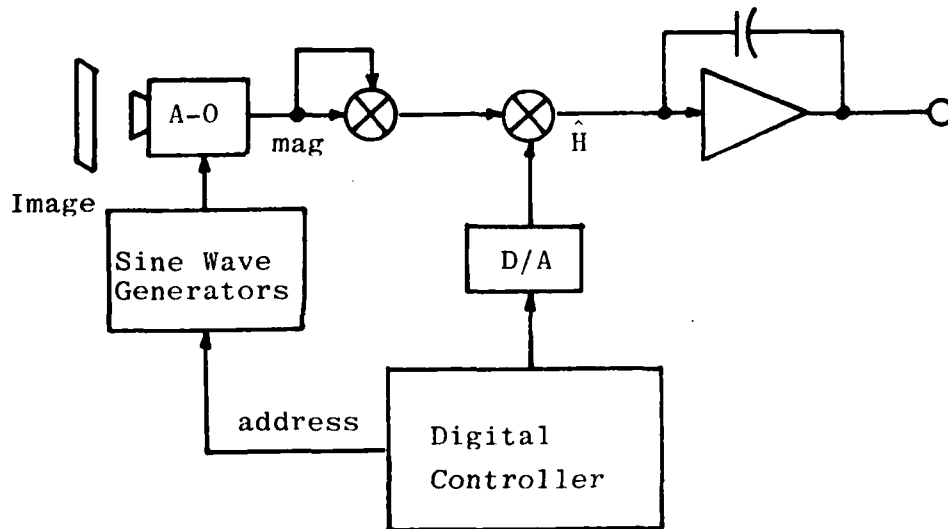


Figure 16 - FF\* Matched Filter : A-0 Implementation

The simplification in implementation when  $F$  is replaced by  $FF^*$  is obvious from a comparison of Figure 15 with Figure 16. In the second case no CRT's are required. In addition, one rather than three A-O devices are required. The matched filter for  $SS^*$  has some functional advantages and disadvantages when compared with the matched filter for  $S$ . The advantages are the following. First, there is a dimensionality reduction from input to output. This simplifies the decision processor. (A two-dimensional search is not required.) Second, the matched filter for  $SS^*$  is invariant to translation which also simplifies the decision processor. The disadvantage of this matched filter is that the detection of a signal with transform  $SS^*$  does not imply that the image contained the signal with transform  $S$ . All signals with transforms  $|S|e^{j\phi(\omega_x, \omega_y)}$  will produce the same result at the matched filter output.  $\phi$  is an arbitrary phase function. Hence, this matched filter can be used to correctly detect features with transform  $S$  but will produce false alarms when certain other images are viewed. This disadvantage may be acceptable if the purpose of the feature extractor is to screen photos with a final decision on feature content being made by another means such as human observations.

As a final comment, neither types of matched filters considered here are invariant to feature rotation or scale change.

### 3. Method of Invariant Fourier Signatures

In this section some prefilter algorithms will be developed which compute feature signatures which are invariant to feature translation, rotation and scale. These algorithms are refinements of the matched filter of the image autocorrelation function which was developed in the previous section. Referring to equation (38),  $\hat{H}$  can be thought of as a weight function which is directly proportional to  $SS^*$  and inversely proportional to  $(NN^*)^2$ . The matched filter output at  $\hat{m}(0,0)$  is simply the integral over all spatial frequencies of the

product of this weight function with  $FF^*$ . As a result, all of the information contained in the fine structure of  $FF^*$  is lost and not available to the decision processor. In fact, all the decision processor gets is a single number on which the decision is to be based.

There are some obvious drawbacks to this type of pre-processor. First, this matched filter is rather insensitive to the shape of the transform  $FF^*$ . That is, there may be two dissimilar images  $I_1$  and  $I_2$  with transforms  $F_1$  and  $F_2$  such that

$$F_1 F_1^* \neq F_2 F_2^* \quad (41)$$

and yet

$$\iint F_1 F_1^* \hat{H} d\omega_x d\omega_y = \iint F_2 F_2^* \hat{H} d\omega_x d\omega_y \quad (42)$$

This would lead to additional, undesirable false alarms. A second drawback which has been mentioned is that the matched filter is not invariant to either rotation or scale changes. Hence, to use matched filters for feature extraction a number of filters must be implemented for different angular orientations and scale factors. Matched filter degradation due to rotation and scale mismatch is discussed in reference 19.

Poor sensitivity of the preprocessor transform shape results because the two-dimensional image is reduced to a scalar. Better sensitivity may be obtainable if the preprocessor is modified so that its output is a vector rather than a scalar. A promising approach first taken by Lendaris and Stanley<sup>20</sup> is to replace the double integral in equation (40) with  $n$  line integrals. That is, the product  $FF^* \hat{H}$  is integrated along  $n$  contours. These integrals are then the components of a  $n$ -vector which is the prefilter output. Two of the contours used by Lendaris and Stanley are shown in Figure 17.

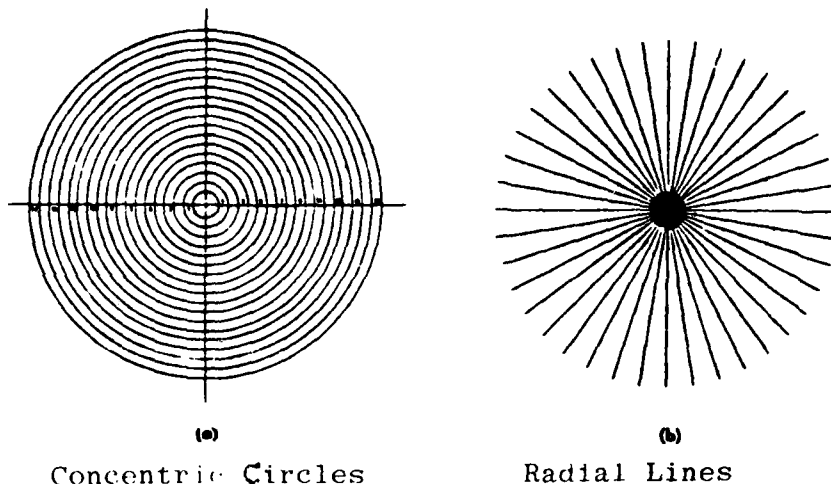


Figure 17 - Integration Contours

In their work they only used the magnitude of the Fourier transform. They did not implement the matched filter  $\hat{H}$ . They were concerned with computing prefilter outputs which are invariant to translation and rotation or to translation and scale but not both. In this section, a more general type of prefilter will be developed which can be invariant to translation, rotation and scale and also allow noise filtering which is analogous to matched filtering. However, to clarify the presentation some algorithms of Lendaris and Stanley will first be developed. These will then be generalized.

To begin, assume that either  $|F|$  or  $FF^*$  is available from an A-0 sensor. Also assume for now that  $F = S$ . That is, there is no noise and the image consists of only the feature to be detected. Now if the feature is translated by  $\Delta x, \Delta y$  then the transform changes to  $Fe^{-j(\Delta x \omega_x + \Delta y \omega_y)}$ . Hence both  $|F|$  and  $FF^*$  do not change and these functions are said to be invariant to translation. We will now consider the two sets of contours of Figure 17 separately.

Consider first the set of concentric circles. Let the  $i$ -th circle have radius  $r_i$ . The difference  $r_i - r_{i-1}$  need not be a constant. There are  $n$  contours. Let  $FF^*$  or  $|F|$  be integrated around any of these contours. Now if the feature were to rotate through angle  $\theta$  then this integral would not change.



This is because  $FF^*$  or  $|F|$  would only rotate around the transform origin by the same angle. This will be true for all contours. Hence, the  $n$ -vector whose components are the  $n$  line integrals is invariant to rotation.

Now consider the second set of contours. Let the  $i$ -th radial line be at an angle  $\theta_i$  from the  $\omega_x$ -axis. The difference  $\theta_i - \theta_{i-1}$  need not be a constant. There are  $n$  contours. Let  $FF^*$  or  $|F|$  be integrated along any of these contours. Now if the feature were to be scaled in size by  $\alpha$  then  $|F|$  and  $FF^*$  would be scaled in size by  $1/\alpha$ . As long as the radial lines extended far enough from the origin so that all the significant energy of the transform will be included in the contour integration for both scaled and unscaled feature, then in both these cases the line integration will remain approximately the same. This will be true for all contours. Hence, the  $n$ -vector whose components are the  $n$  line integrals is (approximately) invariant to scale.

Prefilters based on these algorithms can be implemented easily using A-0 sensors. The sensor spatial frequency address would be incremented along a contour. The sensor output would then be integrated along the contour. A simple integration formula such as the trapezoidal rule

$$\frac{1}{h_s} \left[ \frac{f(x_i) + f(x_{i-1})}{2} \right] \approx \int_a^b f(x) dx \quad (43)$$

where  $h_s = x_i - x_{i-1}$  would probably be adequate. (Note that the Thomson - CSF sensor would be limited to scanning along radial lines. For a discussion refer to Section II D.)

The invariant properties of these algorithms appear to be very useful. They are also noteworthy because only a single Fourier transform is required and, therefore, they can be naturally implemented using a single A-0 device. These algorithms were successfully used by Lendaris and Stanley to detect features in aerial photographs. They have also been used by Pernick, et al, in the screening of cervical cytological

samples<sup>21</sup>. However, these algorithms have some disadvantages.

The first disadvantage is that they do not consider the effect of noise. In the matched filter the output is enhanced by multiplying  $FF^*$  by the weight function  $\hat{H}$ . If the image contains signal plus noise then it is not possible to weight  $FF^*$  by  $\hat{H}$  before integrating along the contours because the feature is assumed to be either rotated or scaled by some unknown factor. Since  $\hat{H}$  is not invariant to these changes in the feature, the required weighting function is not known. The proper place to apply filtering is after the line integrations. Assume that signal and noise are uncorrelated so that equation (37) is satisfied. Since the integral of a sum equals the sum of the integrals then signal and noise will still be additive at the output of the contour integrations. (This implies that  $FF^*$  is used. If  $|F|$  is used, signal and noise are no longer additive.) That is,

$$\phi_i FF^* = \phi_i SS^* + \phi_i NN^* \quad (44)$$

where  $\phi$  is the  $i$ -th contour integral. Hence, the matched filter<sup>1</sup> principle can be applied at the prefilter output after line integration. Let the weight to be applied to the output of the  $i$ -th line integral be  $w_i$ . Then from equation (32)

$$w_i = \frac{\phi_i SS^*}{(\phi_i NN^*)^2} \quad (45)$$

To be specific, consider the case of concentric circles for contours. The feature to be detected has transform  $S$ . Since the algorithm has been shown to be invariant to feature rotation,  $w_i$  will also be invariant to feature rotation. (The noise does not change with feature rotation since feature and noise are assumed uncorrelated.) Hence, the set of weights is a constant  $n$ -vector which only depends on the noise and on the feature but not on the angular orientation of the feature. That is, the weight vector can be computed apriori. This will also be

the case when the contours consist of radial lines. In this case the feature has an arbitrary scale. However, the algorithm is invariant to scale so that  $\phi_{SS}^*$  is constant. Hence the weight vector is again constant and can be computed apriori using equation (45)

In general, let the  $i$ -th component of the measurement vector be  $m_i$  where

$$m_i = \phi_{FF}^* \quad (46)$$

then the prefilter output  $n$ -vector is given by  $V$  where

$$V^T = [w_1 m_1, w_2 m_2, \dots, w_n m_n] \quad (47)$$

Given that matched filtering can be applied to these algorithms they still suffer a disadvantage in the case that the feature may have arbitrary position, rotation and scale. These algorithms are invariant to translation and rotation or translation and scale but not to all three. If the feature is present but suffers all three changes with respect to the reference feature with spectrum  $S$  then the output  $n$ -vector will not correspond to the reference output and detection may not be possible. It is, however, possible to generalize these algorithms so that the prefilter can be made invariant to all three feature changes.

To develop the more general algorithm, begin with the image  $I(x,y)$  with transform  $F(\omega_x, \omega_y)$ . Form either  $|F|$  or  $FF^*$  which has been shown to be invariant to translation. (If matched filtering is to be applied later then  $FF^*$  should be used.) Then form the function

$$G(\rho, \theta) = |F(e^{\rho} \cos \theta, e^{\rho} \sin \theta)| \quad (48)$$

or if  $FF^*$  was formed,

$$G(\rho, \theta) = F(e^{\rho} \cos \theta, e^{\rho} \sin \theta) F^*(e^{\rho} \cos \theta, e^{\rho} \sin \theta) \quad (49)$$

That is,  $G$  is  $|F|$  or  $FF^*$  distorted exponentially in radius and expressed in polar coordinates. The function  $G$  is periodic in  $\theta$  with period  $\pi$ . Assume now that the image consists of signal

with no noise. Consider now the change in  $G$  when the signal or feature in  $K(x,y)$  suffers an arbitrary rotation and scale change. Let  $(r,\gamma)$  be the location of an arbitrary component of  $|F|$  or  $FF^*$  before rotation and scale change. The location of the corresponding component in  $G$  is  $(\ln r, \gamma)$ . After rotation of the feature through angle  $\phi$  and scaling by  $\alpha$  the component in  $|F|$  or  $FF^*$  will move to  $(r/\alpha, \gamma+\phi)$ . The location of the corresponding component in  $G$  is  $(\ln r - \ln \alpha, (\gamma+\phi)_{\text{mod } \pi})$ . That is,  $G$  will be translated by  $-\ln \alpha$  along the  $\rho$  direction and by  $\phi$  along the  $\theta$  direction (modulo  $\pi$ ). Since  $|F|$  or  $FF^*$  is invariant to translation, any combination of translation, rotation and scale change of the feature will result in only a translation in  $G$ . ( $G$  will also suffer a gain change to  $|\alpha|^{-2}G$  but this is not important.)

Since the only change in  $G$  is a translation, by taking a second two-dimensional Fourier transform, this time of  $G(\rho, \theta)$ , and then forming the magnitude or magnitude squared of this second transform, a function is formed which is invariant to translation in  $G$ . Hence, this last function is invariant to translation, rotation and scale of the feature. This function is

$$H(\omega_\rho, \omega_\theta) = \int_{\rho_{\min}}^{\rho_{\max}} \int_{-\pi/2}^{\pi/2} G(\rho, \theta) e^{-j(\omega_\rho \rho + \omega_\theta \theta)} d\rho d\theta \quad (50)$$

The functions  $|H|$  and  $HH^*$  are invariant to translation of  $G$ . Strictly speaking, this is true only along the lines parallel to the  $\omega_\rho$ -axis defined by

$$\omega_\theta = 2n; n \text{ an integer} \quad (51)$$

This is because  $G$  is periodic in  $\theta$  and should be expanded in a Fourier series rather than a Fourier integral in the  $\theta$ -direction. The Fourier integral evaluated on the above lines reduces to the Fourier series.

Although  $|H(\omega_\rho, n/\pi)|$  or  $H(\omega_\rho, n/\pi)H^*(\omega_\rho, n/\pi)$  is a two-dimensional invariant function it is probably not necessary (and certainly not desirable from a computational standpoint) to compute this function. To develop more easily computable

invariant signatures, recourse is made to the Fourier transform projection theorem. Let  $P_\phi[G(\rho, \theta)]$  be the projection of  $G$  onto a line at angle  $\phi$  from the  $\rho$ -axis. The projection theorem states that the one-dimensional Fourier transform of  $P_\phi[G]$  equals  $H(\omega_\rho, \omega_\theta)$  restricted to a line through the origin of  $(\omega_\rho, \omega_\theta)$  - space and at an angle  $\phi$  to the  $\omega_\rho$ -axis<sup>22</sup>. This line is shown dashed in Figure 18. Since  $G(\rho, \theta)$  is periodic in  $\theta$ ,  $P_\phi[G]$  will also be periodic for  $\phi \neq 0$ . Hence,  $P_\phi[G]$  should be expanded in a Fourier series rather than a Fourier transform. Let  $F\{P_\phi[G]\}$  be the Fourier series expansion of  $P_\phi[G]$ . The terms of this series equal  $H(\omega_\rho, \omega_\theta)$  evaluated at the intersection of the dashed line with the horizontal lines  $\omega_\theta = 2n$ .

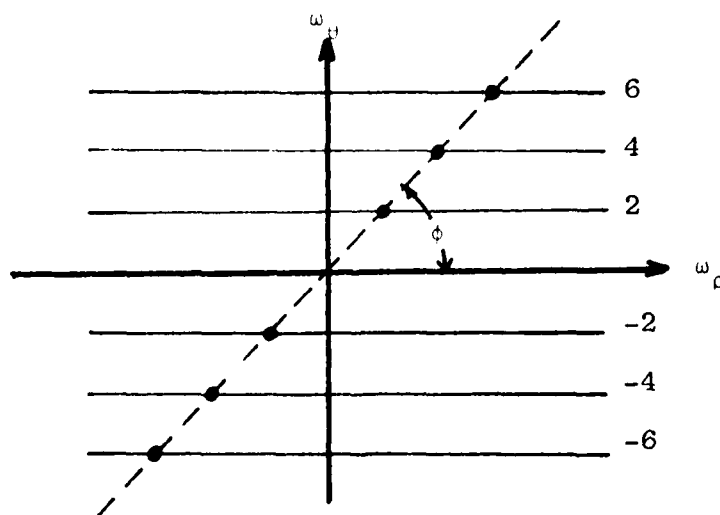


Figure 18 -  $(\omega_\rho, \omega_\theta)$  - Space and  $F\{P_\phi\}$  Line

Since  $|H|$  and  $HH^*$  are invariant, so is  $|F\{P_\phi\}|$  and  $F\{P_\phi\}F^*\{P_\phi\}$ . Now  $F\{P_\phi\}$  contains only some of the information contained in  $H$ . It is conjectured that by properly choosing  $\phi$  values experimentally, a set of invariant signatures could be developed which contain most of the useful information in  $H$ . The signature or signatures to be used may depend on the set of features to be recognized.

Because of the way this development has been carried out, it may not be clear how the invariant signatures are to be computed from the first Fourier transform  $F(\omega_x, \omega_y)$ . This can be clarified with the aid of Figures 19 and 20. In the following it will be assumed that all processing after computation of the Fourier transform of the image will be done digitally. Hence, continuous functions must be sampled to get discrete samples for digital processing.

Figure 19 shows the domain of the pertinent part of  $F$  in  $(\omega_x, \omega_y)$  - space.  $F$  has been restricted to spatial frequencies  $\omega_x, \omega_y$  where  $\rho_{\min} \leq \sqrt{\omega_x^2 + \omega_y^2} \leq \rho_{\max}$ . Because of the mapping  $\rho = e^p$ ,  $F$  will be only sampled on the concentric circles shown in Figure 19. There are  $n$  circles and their radii are  $e^k, e^{2k}, e^{3k}, \dots, e^{nk}$  where

$$k = \ln \rho_{\min} \quad (52)$$

$$\rho_{\max} = e^{nk} \quad (53)$$

Because of the mapping from  $F$  to  $G$ , these circles in  $(\omega_x, \omega_y)$  -space will map to the dashed, equally spaced vertical lines shown in Figure 20 in  $(\rho, \theta)$  - space. Since the image is a real intensity function

$$F(\omega_x, \omega_y) = F(-\omega_x, -\omega_y) \quad (54)$$

so that  $G$  is periodic in the  $\theta$ -direction with period  $\pi$ .

Consider now the formation of the projection  $P_\phi[G]$ .  $P_\phi$  is a function of a single variable  $\mu$  where

$$\mu = \rho \csc \theta \quad (55)$$

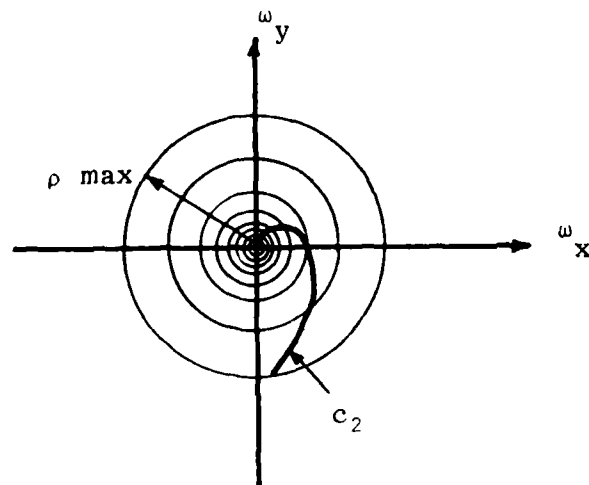


Figure 19 -  $(\omega_x, \omega_y)$  - Space Contours

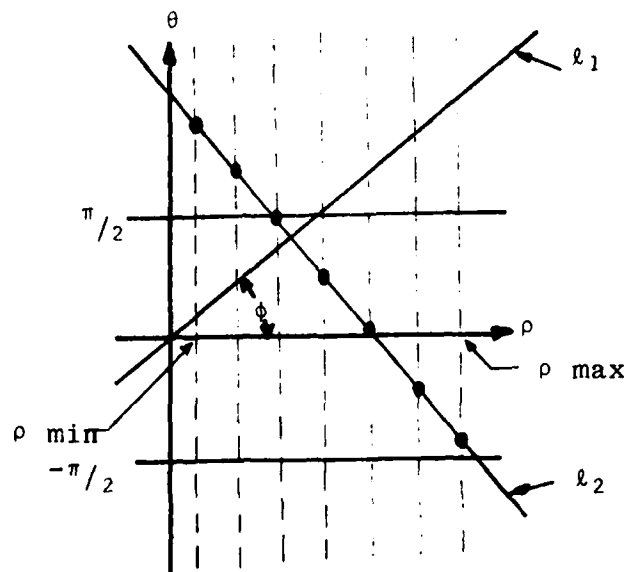


Figure 20 -  $(\rho, \theta)$  - Space Contours

The domain of  $P_\phi$  is the line through the origin of  $(\rho, \theta)$  - space at angle  $\theta$  to the  $\rho$ -axis. This line is labeled  $\ell_1$  in Figure 20.  $P_\phi$  will only be computed at a discrete set of points. Let  $\mu_i$  be one of these points. To determine  $P_\phi(\mu_i)$  a straight line is drawn which passes through the point  $\mu_i$  and is at right angles to  $\ell_1$ . This line is labeled  $\ell_2$  in Figure 20. Now, to find  $P_\phi(\mu_i)$ , simply integrate  $G$  along  $\ell_2$ . Since  $G$  is only available along the vertical, dashed lines of Figure 20, a numerical integration is performed using sample points which are the intersection of  $\ell_2$  with the vertical dashed lines. The values of  $G$  at this set of points is simply the function  $|F|$  or  $FF^*$  evaluated at a set of points which is defined by the intersection of the  $n$ , concentric circles in Figure 19 with the curve which is the mapping of  $\ell_2$  from  $(\rho, \theta)$  - space to  $(\omega_x, \omega_y)$  - space. This curve is labeled  $c_2$  in Figure 19. Note that  $c_2$  is a spiral. This is always the case unless  $\phi = 0^\circ$  or  $\phi = 90^\circ$ . If  $\phi = 0^\circ$  then  $\ell_2$  is parallel to the vertical, dashed lines in Figure 20. In that case  $\mu_i$  is chosen so that  $\ell_2$  coincides with one of these vertical lines. That is,  $c_2$  will be a concentric circle. This implies the  $|F|$  or  $FF^*$  be integrated around a circle in order to compute  $P_\phi(\mu_i)$ . If  $\phi = 90^\circ$  then  $\ell_2$  is parallel to the  $\rho$ -axis. In  $(\omega_x, \omega_y)$  - space the corresponding  $c_2$  will be a radial line. The intersection of this line with the  $n$ , concentric circles defines the sample points used to compute the numerical integration. These special cases where  $\phi = 0^\circ$  or  $90^\circ$  result in contours which are similar to the contours of Figure 17. The only difference is the exponentially distorted spacing of circles. For any other value of  $0^\circ < \phi < 90^\circ$  the integration contours are spirals. Hence, this algorithm is more general than the previously considered one.

Now for whatever  $\phi$  angle used, the above procedure is repeated for each of the sample points  $\mu_i$ . In this way,  $P_\phi[G]$  is computed at a set of sample values. To increase efficiency



of the remaining computation, the number of sample values of  $P_\phi$  is chosen to be a power of two. Now recall that the one-dimensional Fourier series of  $P_\phi[G]$  equals  $H(\omega_\rho, \omega_\theta)$  restricted to a line through the origin of  $(\omega_\rho, \omega_\theta)$ -space and at an angle  $\phi$  to the  $\omega_\rho$ -axis. Hence, the final step in the algorithm is to compute the Fourier series of  $P_\phi$ . Since  $P_\phi$  is only computed at a set of equally spaced samples,  $\nu_i$ , a discrete Fourier transform (DFT) rather than a continuous Fourier series is computed. As long as the sample spacing is sufficiently fine to prevent significant aliasing, the substitution of the discrete Fourier transform for the Fourier series is acceptable. Since  $P_\phi$  has been sampled at  $2^m$  points, the DFT can be computed using an FFT algorithm.

In the above development, the functions  $F$ ,  $G$  and  $H$  have been defined. However, in the actual calculations of the invariant signature  $F\{P_\phi[G]\}$  only the function  $F$  need be considered. In review, to compute the signature the following steps are required:

- 1) A set of radii defined by

$$r_i = e^{ik} \quad (56)$$

for some constant  $k$  is chosen. These radii define the  $n$ , concentric circles. (For the special case that  $\phi = 0^\circ$ ,  $n$  is chosen to be a power of two.)

- 2) A projection angle  $0^\circ \leq \phi \leq 90^\circ$  is chosen.
- 3) An integer  $m$  is chosen to define the FFT length as  $2^m$ .
- 4) If  $\phi < 90^\circ$  then the following step is performed:  
For each  $0 \leq i \leq 2^m - 1$   $|F|$  or  $FF^*$  is integrated along the contour  $c_2$  using samples with (polar) coordinates

$$(e^{ik}, (l-1)\phi + i\pi/2^m) \quad 1 \leq l \leq n \quad (57)$$

The result is a vector of length  $2^m$  with each element of the vector corresponding to a difference value of  $i$ .

- 5) If, instead,  $\phi = 90^\circ$  then the following step is performed:  
For each  $0 \leq i \leq n$   $|F|$  or  $FF^*$  is integrated around each of the  $n$ ,

concentric circles. A discrete integration is performed by only sampling  $F$  discretely around each circle.  $n$  is chosen to be  $2^m$  so that the resulting vector is of length  $2^m$ . Each element corresponds to a different value of  $i$ .

6) An FFT is performed on the  $2^m$ -vector.

7) Either the magnitude or the magnitude squared of the FFT output samples are computed.

The resulting  $2^m$ -vector of real numbers is the desired signature. The above series of steps can be repeated for other values of  $\phi$ , if so desired. The above algorithm will be termed the method of invariant Fourier signatures (IFS). Elements of this algorithm are developed in a series of papers <sup>23, 24, 25, 26, 27</sup>. The procedure of exponentially distorting a function and then computing its Fourier transform can be shown to be equivalent to the Mellin transform<sup>23</sup>.

$$MT(\nu) = \int_0^\infty f(x) x^{-j\nu-1} dx \quad (58)$$

In the above development the question of noise corruption was not considered. The invariant signatures which can be computed from the algorithm are only invariant in the noise-free case. We have already considered the application of the matched filter to the output of the prefilter defined by equation (47). Consider now the extension of this development to the more general prefilters developed above.

To begin, assume that the image  $I(x,y)$  consists of signal plus noise. That is,  $I$  is defined by equation (30). Assume also that signal and noise are uncorrelated so that equation (37) holds. Now let  $\phi_i$  be the result of integrating  $FF^*$  along the  $i$ -th contour which may be a radial line, circle or spiral. Then with the above assumptions equation (44) still holds. That is, the exponential distortion of the radius does not effect this result. Define signal and noise to be  $\tilde{S}(\nu)$  and  $\tilde{n}(\nu)$  where

$$\tilde{S}(\nu_i) = \phi_i SS^* - \frac{1}{n} \sum_{k=1}^n \phi_k SS^* \quad (59)$$

$$\tilde{n}(\mu_i) = \oint_i NN^* - \frac{1}{n} \sum_{k=1}^n \oint_k NN^* \quad (60)$$

That is  $\tilde{s}(\mu_i)$  equals the result of the line integral at  $\mu_i$  with the mean value over all  $i$  removed. The same applies to  $\tilde{n}(\mu_i)$ . (Removal of the mean has not been considered to this point.) Define the transforms of  $\tilde{s}$  and  $\tilde{n}$  to be  $\hat{S}$  and  $\hat{N}$  respectively. That is

$$\hat{S} = F\{\tilde{s}\} \quad (61)$$

$$\hat{N} = F\{\tilde{n}\} \quad (62)$$

and define

$$\hat{F}\{P_\phi[G]\} = \hat{S} + \hat{N} \quad (63)$$

( $\hat{F}\{P_\phi\}$  is simply  $F\{P_\phi\}$  with the zero frequency term set to zero.) Now matched filtering can be applied since signal and noise are additive which follows from equation (44). However, to avoid another transform, we make the further assumption that  $\hat{S}$  and  $\hat{N}$  are uncorrelated so that

$$\hat{F}\{P_\phi\} \hat{F}^*\{P_\phi\} = \hat{S}\hat{S}^* + \hat{N}\hat{N}^* \quad (64)$$

The removal of the means was necessary to make this assumption a possibility. The corresponding matched filter weight function is

$$\hat{H} = \frac{\hat{S}\hat{S}^*}{(\hat{N}\hat{N}^*)^2} \quad (65)$$

The matched filter output is

$$\tilde{m}(o) = \int \hat{F}\{P_\phi\} \hat{F}^*\{P_\phi\} \hat{H} dv \quad (66)$$

where  $v$  is the frequency variable.

Notice that the result is a scalar rather than a  $2^m$  vector. What has happened is that by adding matched filtering, the decision processor algorithm has been included with the pre-processor algorithm. The only additional computation required is to compare  $\tilde{m}(o)$  with a threshold. Note the form of  $\hat{H}$ . It is the output which would occur if only the signal (feature) were present divided by the square of the output due to the

noise alone. Hence, where a large noise contribution was expected the output would be de-emphasized.  $\hat{m}(o)$  is simply the inner product of the prefilter output with the matched filter weight.

Finally, consider the implementation of any of these algorithms using A-0 devices. A block diagram is shown in Figure 21.

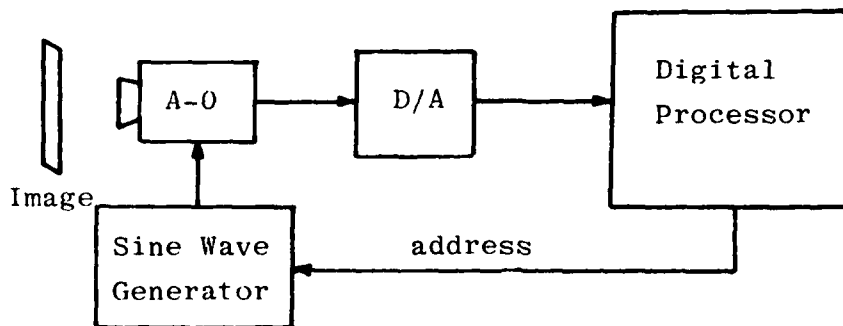


Figure 21 - Invariant Fourier Signatures: Implementation

The block diagram is rather simple. The A-0 device is used to produce components of the two-dimensional Fourier transform of the image. The sample points are controlled by addressing the sensor. The remainder of the processing takes place in the digital processor. The amount of data memory required is  $2^{m+1}$  words to hold the complex  $2^m$  vector. Spatial frequency addresses can either be computed using a cordic algorithm<sup>28</sup> or precomputed and stored in ROM. It can be seen that these algorithms are well-suited for implementation with A-0 devices since the two-dimensional image information is immediately reduced to a much smaller set of numbers which can be post-processed in the digital processor.

In summary, a flexible class of algorithms has been developed which can be used as a feature extractor prefilter. These algorithms can be translation, rotation and scale

invariant. This is strictly true only in the case where no noise is present. However, matched filtering can be used to minimize the noise at the prefilter output. The parameter  $\phi$  can be used to optimize the prefilter for the particular feature or class of features to be identified in the image set. The parameter  $\phi$  can be arbitrarily chosen because A-0 devices can be addressed at arbitrary spatial frequencies. Finally, these algorithms are well-suited for implementation using an A-0 device for a Fourier transformer and a mini-computer as a post-processor.

#### 4. Method of Invariant Moment Signatures

The method of moments is used in a number of disciplines. The utility of moments in feature extraction was first pointed out by Hu<sup>12</sup>. It can be shown that algebraic combinations of image moments are invariant to translation, rotation and scale. This section is a discussion of image moments and how they can be computed using A-0 devices.

The two-dimensional  $(p+q)$ -th moments of an image intensity function  $I(x,y)$  are defined by

$$m_{p,q} = \iint x^p y^q I(x,y) dx dy \quad (67)$$

It can be shown that the double sequence of moments  $\{m_{p,q}\}$  is uniquely determined by  $I(x,y)$  and conversely,  $I(x,y)$  is uniquely determined by  $\{m_{p,q}\}$  as long as  $I(x,y)$  satisfies certain conditions which are always met with real images<sup>12</sup>. A method of reconstructing  $I(x,y)$  from  $\{m_{p,q}\}$  is given in reference 13.

The purpose of computing image moments is two-fold. First, the invariant functions of moments are desirable for feature extraction. Second, the possibility exists of replacing the image, which contains a large amount of information represented by its pixel values, with a much smaller set of moments  $\{m_{p,q}\}$  such that  $p \leq p_{\max}$  and  $q \leq q_{\max}$ . This is analogous to computing the two-dimensional Fourier

transform for the purpose of reducing the image information to a much smaller set of spatial frequencies.

Consider now the central moments which are defined by

$$\mu_{p,q} = \iint (x-\bar{x})^p (y-\bar{y})^q I(x,y) d(x-\bar{x}) d(y-\bar{y}) \quad (68)$$

where

$$\bar{x} = m_{1,0}/m_{0,0} \quad (69)$$

$$\bar{y} = m_{0,1}/m_{0,0} \quad (70)$$

It is well known that the central moments are invariant under translation of coordinates. Hence, the central moments are translation invariant. The central moments can be expressed in terms of the ordinary moments. For example

$$\mu_{0,0} = m_{0,0} \quad (71)$$

$$\mu_{1,0} = \mu_{0,1} = 0 \quad (72)$$

$$\mu_{2,0} = m_{2,0} - m_{0,0} \bar{x}^2 \quad (73)$$

$$\mu_{0,2} = m_{0,2} - m_{0,0} \bar{y}^2 \quad (74)$$

$$\mu_{1,1} = m_{1,1} - m_{0,0} \bar{x}\bar{y} \quad (75)$$

Similar expressions are easily derived for all higher order moments with the use of equation (68). If, for example, the ordinary moments could be computed using an A-0 device then the central moments could be formed in a digital post processor.

Consider the information contained in the first few moments

$$\{\mu_{0,0}, \mu_{1,0}, \mu_{0,1}, \mu_{2,0}, \mu_{0,2}, \mu_{1,1}\}$$

$$\mu_{0,0} = m_{0,0} = \iint I(x,y) dx dy \quad (76)$$

represents the total image power. Both  $\mu_{1,0}$  and  $\mu_{0,1}$  are zero. However, the ordinary moments

$$m_{1,0} = \iint x I(x,y) dx dy \quad (77)$$

$$m_{0,1} = \iint y I(x,y) dx dy \quad (78)$$

locate the image centroid which is  $(\bar{x}, \bar{y})$  where  $\bar{x}$  and  $\bar{y}$  have already been defined.

The above set of moments characterize the size, gross shape and orientation of the image. If only these moments are considered then the image moments are identical to the moments of an image consisting of the ellipse shown in Figure 22<sup>13</sup>. The parameters of this ellipse are given by

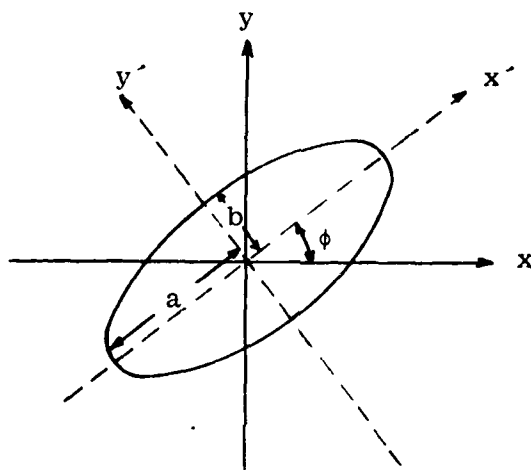


Figure 22 - Equivalent Image Ellipse

$$a = \left( \frac{\mu_{2,0} + \mu_{0,2} + [(\mu_{2,0} - \mu_{0,2})^2 + 4\mu_{1,1}^2]^{1/2}}{\mu_{0,0}^{1/2}} \right)^{1/2} \quad (79)$$

$$b = \left( \frac{\mu_{2,0} + \mu_{0,2} - [(\mu_{2,0} - \mu_{0,2})^2 + 4\mu_{1,1}^2]^{1/2}}{\mu_{0,0}^{1/2}} \right)^{1/2} \quad (80)$$

$$\phi = (1/2) \tan^{-1} \left( \frac{2\mu_{1,1}}{\mu_{2,0} - \mu_{0,2}} \right) \quad (81)$$

Hence, if only these moments are known then only very general information about the image shape is available. However, even in this case enough information may be available to achieve some feature extraction. For example, it should be possible to distinguish long, thin features such as roads, airport runways

and shorelines from compact features such as buildings and vehicles. Long, thin objects are characterized by  $a \gg b$  or  $b \gg a$  while compact features are characterized by  $a \approx b$ . Rotating the image results in change in the angle  $\phi$ . However, image rotation will not change  $a$  or  $b$ . Hence,  $a$  and  $b$  are invariant to rotation. Because they are formed from central moments they are also invariant to translation. And, finally, because of the  $\mu_{0,0}$  term in the denominator they are invariant to intensity changes.

Since  $a$  and  $b$  are invariant to translation and rotation so are

$$a^2 + b^2 = \frac{2(\mu_{2,0} + \mu_{0,2})}{\mu_{0,0}^{1/2}} \quad (82)$$

$$a^2 - b^2 = \frac{2[(\mu_{2,0} - \mu_{0,2})^2 + 4\mu_{1,1}^2]^{1/2}}{\mu_{0,0}^{1/2}} \quad (83)$$

These last two moment invariants were used by Hu to distinguish between letters of the alphabet<sup>12</sup>.

Moments can be modified to be invariant to scale. Consider the scale change  $x' = \alpha x$ ,  $y' = \alpha y$ . Then it is straightforward to show that the unprimed central moments  $\mu_{p,q}$  will change to the primed central moments  $\mu'_{p,q}$  where

$$\mu'_{p,q} = \alpha^{(p+q+2)} \mu_{p,q} \quad (84)$$

To achieve invariance to scale first make use of the relationship

$$\mu'_{0,0} = \alpha^2 \mu_{0,0} \quad (85)$$

so that

$$\alpha^{(p+q+2)} = \left( \frac{\mu'_{0,0}}{\mu_{0,0}} \right)^{1+(p+q)/2} \quad (86)$$

Substituting this expression into equation (84) and separating primed from unprimed terms yields

$$\frac{\mu'_{p,q}}{(\mu'_{0,0})^{1+(p+q)/2}} = \frac{\mu_{p,q}}{\mu_{0,0}^{1+(p+q)/2}} \quad (87)$$



Hence,

$$\tilde{\mu}_{p,q} = \mu_{p,q} / \mu_{0,0}^{1+(p+q)/2} \quad (88)$$

is invariant to scale. The moments  $\tilde{\mu}_{p,q}$  will be termed scale-normalized moments.

The direct substitution of  $\tilde{\mu}$  moments for  $\mu$  moments in the expressions for  $a, b, a^2+b^2$  and  $a^2-b^2$  yields invariants  $\tilde{a}, \tilde{b}, \tilde{a}^2+\tilde{b}^2$  and  $\tilde{a}^2-\tilde{b}^2$  respectively. These invariants are translation, rotation and scale invariant. Note, although  $a, b$  and  $a^2-b^2$  are invariant to intensity variation,  $\tilde{a}, \tilde{b}$  and  $\tilde{a}^2+\tilde{b}^2$  are not. To see this, let  $I(x,y)$  change to  $kI(x,y)$  then  $a, b$  and  $a^2-b^2$  do not change while  $\tilde{a}$  and  $\tilde{b}$  change to  $k^{-1/2}\tilde{a}$  and  $k^{-1/2}\tilde{b}$  and  $\tilde{a}^2+\tilde{b}^2$  changes to  $k^{-1}(\tilde{a}^2+\tilde{b}^2)$ . This is not necessarily a problem, it should be kept in mind when applying these formulas, however. It should be noted that the normalization given in equation (88) is not the only one possible. Another scale-normalized set of moments is given by <sup>9,13</sup>

$$\hat{\mu}_{p,q} = \mu_{p,q} / (\mu_{2,0} + \mu_{0,2})^{(2+p+q)/4} \quad (89)$$

Up to this point moments and central moments have been defined. A few scalars have been given which are invariant to translation, rotation and scale. These results can be generalized. In references 12 and 13 it is shown that by including higher order moments, additional invariant scalars can be derived which are translation, rotation and scale invariant. Higher order moments contain more information about the image. In his paper<sup>12</sup> Hu was able to differentiate long, slender letters such as I or L from compact letters such as N or M using only  $a^2-b^2$ . However, he was not able to differentiate between letters which have about the same shape such as the pairs (W,M), (E,F) or (B,R). In a more recent paper<sup>13</sup> Teague has shown that moments up to at least the 11th or 12th order are needed to distinguish an E from an F.

Assume now that a sufficiently high number of moments are available so that  $m$ , invariant scalar functions of these moments could be computed. Then the  $m$ -vector of these scalars could serve as an invariant feature signature. This  $m$ -vector would then be the output of the prefilter. This approach has been used to identify aircraft, for example.<sup>9</sup>

Consider now the computation of image moments using A-0 devices. Two approaches to this problem will be considered. They are the following:

1. Computing image moments from the two-dimensional Fourier transform. The Fourier transform is computed by the A-0 device.

2. Computing image moments directly by modifying the A-0 device input signals  $g(x)$  and  $h(x)$  in equation 1.

First, consider the approach requiring the Fourier transform. Some of the results in this development will be applicable to the second approach. Given that the two-dimensional Fourier transform is available, it is possible to compute image moments from samples of the Fourier transform. Two methods of accomplishing this will be considered.

The first method can be developed as follows. Start with the Fourier transform of the image.

$$F(\omega_x, \omega_y) = \iint I(x, y) e^{-j(\omega_x x + \omega_y y)} dx dy \quad (90)$$

Now expand the exponential in a power series and integrate term by term. The result is

$$F(\omega_x, \omega_y) = \sum_{p=0}^{\infty} \sum_{q=0}^{\infty} \frac{(-j)^{p+q}}{p!q!} m_{p,q} \omega_x^p \omega_y^q \quad (91)$$

so that

$$m_{p,q} = (-j)^{-(p+q)} \left[ \left( \frac{\partial}{\partial \omega_x} \right)^p \left( \frac{\partial}{\partial \omega_y} \right)^q F(\omega_x, \omega_y) \right]_{\omega_x = \omega_y = 0} \quad (92)$$

Hence,  $m_{p,q}$  can be derived from the partial derivatives of  $F$  evaluated at the origin. This approach has been considered by Teague<sup>29</sup>. Since  $F(\omega_x, \omega_y)$  and not its partial derivatives

are available, the partial derivatives must be approximated by numerical differentiation. Partial derivatives are replaced by finite differences

$$\left(\frac{\partial}{\partial \omega_x}\right)^p \left(\frac{\partial}{\partial \omega_y}\right)^q F(\omega_x, \omega_y) \approx (\Delta \omega_x)^p (\Delta \omega_y)^q F(\omega_x, \omega_y) \quad (93)$$

where

$$\Delta \omega_x F(\omega_x, \omega_y) = \frac{F(\omega_x + h, \omega_y) - F(\omega_x - h, \omega_y)}{2h} \quad (94)$$

and

$$\Delta \omega_y F(\omega_x, \omega_y) = \frac{F(\omega_x, \omega_y + h) - F(\omega_x, \omega_y - h)}{2h} \quad (95)$$

It can be shown<sup>29</sup> that in order to measure all moments of order  $n$  or less (i.e.,  $p+q \leq n$ ) requires  $2n^2 + 2n + 1$  distinct samples of the transform. The location of these samples is shown in Figure 23. In order to compute moments of order  $n$  or less, the required samples (which are the dots) are all located on or within the square labeled  $n$ .

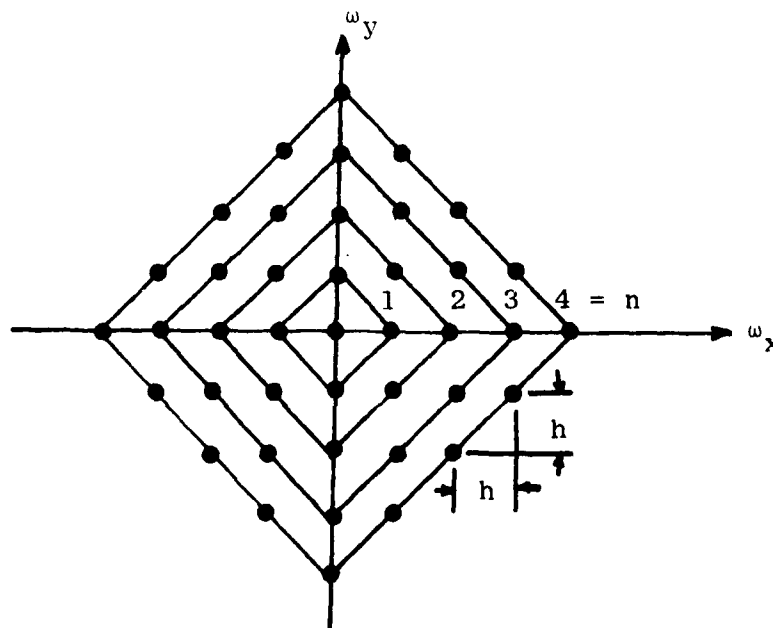


Figure 23 - Transform Sampling: Finite Differences

The only control on accuracy is the step size  $h$ . For high accuracy,  $h$  may be so small that all the samples are restricted to the area of the transform occupied by the d.c. sinc function. (The Fourier transform of an image will always contain a term centered at  $\omega_x = \omega_y = 0$  due to the average light intensity of the image. This term is a square, two-dimensional sinc function because the A-0 sensor is assumed to have a square aperture.) If this is the case then higher frequency components of the transform will not be utilized in the computation of the moments. While it is true that if the Fourier transform were known with infinite precision then decreasing  $h$  will always result in greater accuracy. The exact Fourier transform is an analytic function and so the entire transform can be reconstructed from the partial derivatives evaluated at  $\omega_x = \omega_y = 0$ . However, in practical A-0 sensors, greater measurement accuracy can be achieved by utilizing Fourier transform data over the entire region of Fourier space which is available. The small effect of a high spatial frequency component on the partial derivatives at the origin will be error-prone whereas an accurate measurement at the high spatial frequency is possible. Hence, this method of computing moments does not appear well-suited for implementation using A-0 devices since there is no practical means available for controlling accuracy of computation.

Because of this problem an alternative method was developed under this contract. This method also uses samples of the Fourier transform. However, the sample spacing is fixed and the accuracy of the method is adjusted by adjusting the number of samples. In this method, higher accuracy requires higher spatial frequency components. The method seems well-suited for A-0 device implementation and it will now be developed.

The starting point in this development is the defining

equation for two-dimensional moments, equation (67). Now the A-0 device has a finite aperture. This aperture will be considered square. Because of this, the weight functions  $x^p, y^q$  are only required to have this form within the device aperture. Let these functions be periodically extended beyond the device aperture. This is shown for the one-dimensional case in Figure 24. The extension to two dimensions is straightforward. The device aperture is considered to be centered at  $x = 0$  and to extend from  $x = -1$  to  $x = +1$ . Four cases are shown in the figure. These are : a)  $p$  even and period = 2; b)  $p$  even and period = 4; c)  $p$  odd and period = 2; d)  $p$  odd and period = 8. The periodic extension of  $x^p$  is indicated with the notation  $[x^p]$ . The periodic extension of  $y^q$  is  $[y^q]$ . Over a single period, centered around  $x = 0$ , the functional definition of  $[x^p]$  for the four cases shown in the figure are as follows:

$p$  even and

$$\text{period} = 2: [x^p] = x^p; -1 \leq x \leq 1 \quad (96)$$

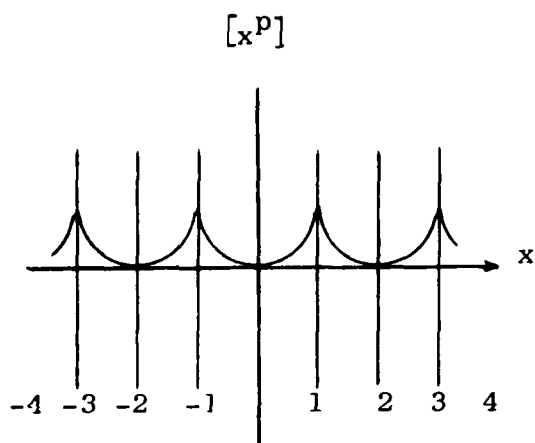
$p$  even and

$$\text{period} = 4: [x^p] = \begin{cases} 2-(x+2)^p; & -2 \leq x \leq -1 \\ x^p; & -1 \leq x \leq +1 \\ 2-(x-2)^p; & +1 \leq x \leq +2 \end{cases} \quad (97)$$

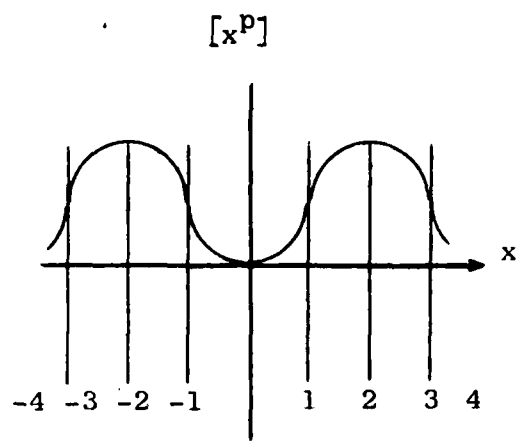
$p$  odd and

$$\text{period} = 2: [x^p] = x^p; -1 \leq x \leq 1 \quad (98)$$

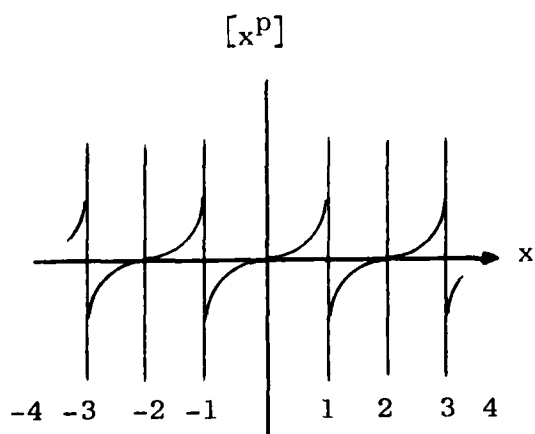
$$\begin{array}{l} p \text{ odd and} \\ \text{period} = 8: [x^p] = \end{array} \left\{ \begin{array}{ll} -(x+4)^p & ; -4 \leq x \leq -3 \\ -2-(x+2)^p & ; -3 \leq x \leq -2 \\ -2+(x+2)^p & ; -2 \leq x \leq -1 \\ x^p & ; -1 \leq x \leq +1 \\ 2+(x-2)^p & ; +1 \leq x \leq +2 \\ 2-(x-2)^p & ; +2 \leq x \leq +3 \\ -(x-4)^p & ; +3 \leq x \leq +4 \end{array} \right. \quad (99)$$



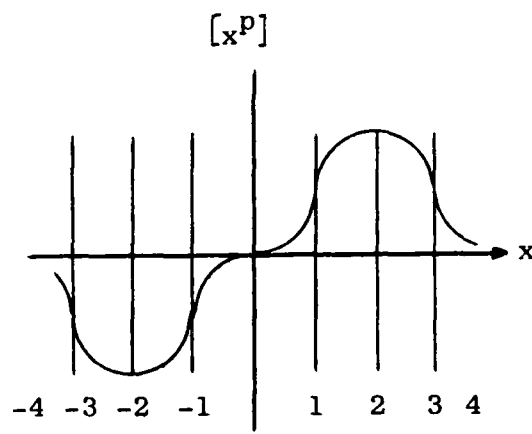
a.  $p$  even and period = 2



b.  $p$  even and period = 4



c.  $p$  odd and period = 2



d.  $p$  odd and period = 8

Figure 24 - Periodic Extensions of  $x^p$

Since  $[x^p]$  is periodic, it can be approximated by a finite sum of complex exponentials. There are a number of ways of achieving such an approximation. For example, the finite sum will be "best" in a mean squares sense if a truncated Fourier series is used.

An alternative method, the method of trigonometric interpolation, will be developed here<sup>30</sup>. This method has the advantage that simple formulas are available for determining the coefficients of the expansion. The method can therefore be easily modified to produce exponential sum approximations to any periodically extended function for which a closed form formula exists without the need for an integration. (An integration is usually required to determine Fourier series coefficients.) The method is, therefore, well-suited for computer generation of the required coefficients. Although the method is very general, the following development will be restricted to the special case of interest, namely the expansion of  $[x^p]$ .

The method has the property that the approximation to  $[x^p]$  will exactly equal  $[x^p]$  on a set of grid points

$$x_i = iL/n; i = 0, \pm 1, \pm 2, \dots, \pm n \quad (100)$$

where  $L$  is one half the period of  $[x^p]$ . The number of points,  $n$ , is a parameter which can be set to achieve the required accuracy.

The approximations to  $[x^p]$  and  $[y^q]$  will be denoted by  $g_p(x)$  and  $h_q(y)$  respectively where

$$g_p(x) = \sum_{k=-n}^n c_k(p) e^{j(k\pi x/L)} \quad (101)$$

$$h_q(y) = \sum_{k=-n}^n c_k(q) e^{j(k\pi y/L)} \quad (102)$$

The notation  $c_k(p)$ ,  $c_k(q)$  for the coefficients is used to indicate that these are functions of the powers  $p$  and  $q$ .

The notation  $\Sigma'$  means

$$\sum_{k=-n}^n a_k = \sum_{k=-n}^n a_k - 1/2(a_n + a_{-n}) \quad (103)$$

Now refer back to Figure 24. The function  $[x^p]$  has different properties depending on whether  $p$  is even or odd and the period 2, 4 or 8. If  $p$  is even and period = 2 then  $[x^p]$  is continuous but the first derivative is discontinuous. If  $p$  is even and period = 4 then both  $[x^p]$  and its first derivative are continuous. If  $p$  is odd and period = 2 then  $[x^p]$  is discontinuous. If  $p$  is odd and period = 8 then both  $[x^p]$  and its derivative are continuous. If either  $[x^p]$  or its derivative is discontinuous at  $x = \pm 1$  the error in the approximation near these points will be rather large unless  $n$  is large. Therefore, it appears better to use the periodic extensions shown in Figure 24 b. and c. These extensions will be exclusively used throughout the remainder of this development. Notice that if  $p$  is odd the period must be twice as long as when  $p$  is even. To account for this,  $L$  in equations (101) and (102) will henceforth be denoted, when necessary, by  $L_e$  or  $L_o$  for the even or odd case respectively. If the normalization of Figure 24 is used then  $L_e = 2$  and  $L_o = 4$ .

The formula for finding  $c_k(p)$  given  $[x^p]$  is given in reference 30. First, express the complex coefficient  $c_k(p)$  in terms of its real and imaginary parts

$$c_k(p) = \frac{\alpha_k(p) - j\beta_k(p)}{2} \quad (104)$$

where

$$\alpha_k(p) = \frac{1}{n} \sum_{\ell=-n}^n g_p(x_\ell) \cos(k\pi x_\ell/L) \quad (105)$$

$$\beta_k(p) = \frac{1}{n} \sum_{\ell=-n}^n g_p(x_\ell) \sin(k\pi x_\ell/L) \quad (106)$$

and  $x_\ell = \ell L/n$ . The formulas for  $c_k(q)$ ,  $\alpha_k(q)$  and  $\beta_k(q)$  are the same except that  $p$  is replaced by  $q$ ,  $x$  is replaced by  $y$



and  $g_p$  is replaced by  $h_q$  everywhere they occur. Since the following development is identical for both  $g_p$  and  $h_q$ , only the case for  $g_p$  will be shown.

The function  $g_p(x)$  equals  $[x^p]$  at all of the  $2n+1$  points  $x_\ell$  (and also at all periodic extensions of these points). There will generally be some error between the mesh points. This will be investigated later.

Now  $g_p$  is either an even or an odd function depending on whether  $p$  is even or odd. For  $p$  even

$$c_k(p) = 1/2 \alpha_k(p) \quad (107)$$

for  $p$  odd

$$c_k(p) = -1/2j \beta_k(p) \quad (108)$$

For  $p$  even or odd  $L_e$  or  $L_o$  is used. To obtain equivalent accuracy, the mesh space should be the same in both cases. Since  $L_o = 2L_e$  the number,  $n$ , of mesh points will be different in each case. Denote them as  $n_e$  and  $n_o$  for even and odd cases respectively. The formula for  $c_k(p)$  and  $p$  even is then given by

$$p \text{ even: } c_k(p) = \frac{1}{2n} \sum_{\ell=-n_e}^{n_e} g_p(x_\ell) \cos(k\pi x_\ell / L_e) \quad (109)$$

$$= \begin{cases} 1; & p = 0 \text{ and } k = 0 \\ 0; & p = 0 \text{ and } k \neq 0 \\ \frac{1}{n} \sum_{\ell=1}^{n_e} g_p(x_\ell) \cos(k\pi x_\ell / L_e); & p > 0 \end{cases} \quad (110)$$

and for  $p$  odd,

$$p \text{ odd: } c_k(p) = -\frac{j}{2n} \sum_{\ell=-n_o}^{n_o} g_p(x_\ell) \sin(k\pi x_\ell / L_o) \quad (111)$$

$$= -\frac{j}{n} \sum_{\ell=1}^{n_o} g_p(x_\ell) \sin(k\pi x_\ell / L_o) \quad (112)$$

where the notation  $\Sigma'$  here means

$$\sum_{\ell=1}^{n'} a_{\ell} = \sum_{\ell=1}^n a_{\ell} - 1/2 a_n \quad (113)$$

If  $p$  is even then  $g_p(x_{\ell})\cos(k\pi x_{\ell}/L_e)$  is an even function. Likewise, if  $p$  is odd then  $g_p(x_{\ell})\sin(k\pi x_{\ell}/L_o)$  is an odd function. Because of this

$$c_k(p) = c_{-k}(p); \quad p \text{ even} \quad (114)$$

$$c_k(p) = -c_{-k}(p); \quad p \text{ odd} \quad (115)$$

These expressions reduce the number of multiplies required to compute  $g_p(x)$  from equation (101) by a factor of two.

There is an additional symmetry present in  $[x^p]$  which reduces the number of multiplies still further. The cases of  $p$  even and odd must be considered separately.

First, consider the case where  $p$  is even and  $p > 0$ . Consider the function

$$\hat{g}_p(x) = g_p(x-1) - 1; \quad p > 0 \text{ and even} \quad (116)$$

in the interval  $-1 \leq x \leq +1$  the function  $\hat{g}_p(x)$  is an odd function while the function  $\cos(k\pi(x-1)/L_e)$  is even in this same interval for  $k$  even and  $k > 0$ . Hence, there will be no contribution to  $c_k(p)$  from the points in this interval when  $k > 0$  and even. An examination of the function

$$\tilde{g}_p(x) = g_p(x+1) - 1; \quad p > 0 \text{ and even} \quad (117)$$

in the interval  $-1 \leq x \leq +1$  leads to the same result. Hence, for  $p > 0$  and even and  $k > 0$  and even  $c_k(p) = 0$ .

Now consider the case the  $p$  is odd. Consider the function

$$\bar{g}_p(x) = g_p(x-2); \quad p \text{ odd} \quad (118)$$

in the interval  $-2 \leq x \leq +2$ . The function  $\bar{g}_p(x)$  is even in this interval while the function  $\sin(k\pi(x-2)/L_o)$  is odd in this same interval for  $k$  even and  $k > 0$ . Hence, there will be no contribution to  $c_k(p)$  from points in this interval when  $k > 0$

and even. An examination of the function

$$\bar{g}_p(x) = g_p(x+2); p \text{ odd} \quad (119)$$

in the interval  $-2 \leq x \leq 2$  leads to the same result. Hence, for  $p$  odd and  $k > 0$  and even,  $c_k(p) = 0$ . In summary:

$$c_k(p) = c_{-k}(p); p \text{ even} \quad (120)$$

$$c_k(p) = c_{-k}(p); p \text{ odd} \quad (121)$$

$$c_k(p) = 0; p \text{ even and } k > 0 \text{ and even} \quad (122)$$

$$c_k(p) = 0; p \text{ odd and } k > 0 \text{ and even} \quad (123)$$

Now consider the generalization to two-dimensions of the results to this point. Let  $[x^p][y^q]$  be the periodic extension of  $x^p y^q$  outside the square with corners  $(-1, -1)$ ,  $(-1, 1)$ ,  $(1, -1)$ ,  $(1, 1)$ . Then  $[x^p][y^q]$  can be approximated by  $g_p(x)h_q(y)$  where

$$g_p(x)h_q(y) = \sum_{i=-n}^n \sum_{k=-n}^n c_i(p)c_k(q)e^{j\pi(ix/L+ky/L)} \quad (124)$$

where the various  $L$ 's and  $n$ 's are to be replaced by  $L_e, L_o, n_e, n_o$  - whichever is appropriate. Now consider the defining equation for  $m_{p,q}$  equation (67). If the limits of integration are restricted to  $-1 \leq x, y \leq 1$  then the kernel  $x^p y^q$  can be replaced by  $[x^p][y^q]$  without effecting the result. That is,

$$m_{p,q} = \int_{-1}^1 \int_{-1}^1 [x^p][y^q] I(x,y) dx dy \quad (125)$$

Now approximate  $[x^p][y^q]$  by  $g_p(x)h_q(y)$

$$m_{p,q} \approx \int_{-1}^1 \int_{-1}^1 g_p(x)h_q(y) I(x,y) dx dy \quad (126)$$

Now substitute the defining trigonometric formulas for  $g_p$  and  $h_q$  and interchange integration and summation. The result is

$$m_{p,q} \approx \sum_{i=-n}^n \sum_{k=-n}^n c_i(p)c_k(q) \int_{-1}^1 \int_{-1}^1 I(x,y) e^{j\pi(ix/L+ky/L)} dx dy \quad (127)$$

$$= \sum_{i=-n}^n \sum_{k=-n}^n c_i(p) c_k(q) F\left(-\frac{i\pi}{L}, -\frac{k\pi}{L}\right) \quad (128)$$

Since  $I(x,y)$  is real  $F(\omega_x, \omega_y) = F^*(-\omega_x, -\omega_y)$

so that

$$m_{p,q} \approx \sum_{i=-n}^n \sum_{k=-n}^n c_i(p) c_k(q) F^*\left(\frac{i\pi}{L}, \frac{k\pi}{L}\right) \quad (129)$$

this is the desired result. It states that if  $x^p y^q$  are approximated over the aperture  $-1 \leq x, y \leq +1$  by the trigonometric series  $g_p(x) h_q(y)$  with coefficients  $c_i(p)$  and  $c_k(q)$  then  $m_{p,q}$  is approximated by the weighted double sum of two-dimensional Fourier transform samples with the same coefficients  $c_i(p)$  and  $c_k(q)$ . Notice that when a larger value of  $n$  is used in order to increase accuracy then the Fourier transform must be sampled over a larger region of Fourier space.

Now, taking into account the symmetries present in the coefficients  $c_i(p)$  and  $c_k(q)$  and in the two-dimensional Fourier transform equation (129) can be simplified. First of all, define the notation

$$\sum_{i=0}^n a_i = \sum_{i=0}^n a_i - 1/2(a_0 + a_n) \quad (130)$$

then the summation approximating  $m_{p,q}$  can be broken into four parts.

$$\begin{aligned} m_{p,q} \approx & \sum_{i=0}^n \sum_{k=0}^n c_i(p) c_k(q) F^*\left(\frac{i\pi}{L}, \frac{k\pi}{L}\right) \\ & + \sum_{i=0}^n \sum_{k=0}^{-n} (\text{same}) + \sum_{i=0}^{-n} \sum_{k=0}^n (\text{same}) + \sum_{i=0}^{-n} \sum_{k=0}^{-n} (\text{same}) \end{aligned} \quad (131)$$

Since  $I(x,y)$  is real  $F(\omega_x, \omega_y) = F^*(-\omega_x, -\omega_y)$  so that

$$\begin{aligned} m_{p,q} \approx & \sum_{i=0}^n \sum_{k=0}^n [c_i(p) c_k(q) F^*\left(\frac{i\pi}{L}, \frac{k\pi}{L}\right) + c_{-i}(p) c_{-k}(q) F\left(\frac{i\pi}{L}, \frac{k\pi}{L}\right)] \\ & + \sum_{i=0}^n \sum_{k=0}^n [c_i(p) c_{-k}(q) F^*\left(\frac{i\pi}{L}, -\frac{k\pi}{L}\right) + c_{-i}(p) c_k(q) F\left(\frac{i\pi}{L}, -\frac{k\pi}{L}\right)] \end{aligned} \quad (132)$$

Now  $F+F^* = 2 \text{ real } (F)$ ,  $F-F^* = 2j \text{ imag}(F)$  and taking advantage of the symmetries equations (120), (121), (122) and (123) the formula for  $m_{p,q}$  can be broken into four pieces for the four cases

1.  $p$  and  $q$  even
2.  $p$  even and  $q$  odd
3.  $p$  odd and  $q$  even
4.  $p$  and  $q$  odd

These cases are the following:

case 1:  $p$  and  $q$  even

$$m_{p,q} = 2 \sum_{i=0}^{n_e} \sum_{k=0}^{n_e} c_i(p) c_k(q) [\text{real } F(i\pi/L_e, k\pi/L_e) + \text{real } F(i\pi/L_e, -k\pi/L_e)] \quad (133)$$

case 2:  $p$  even and  $q$  odd

$$m_{p,q} = 2j \sum_{i=0}^{n_e} \sum_{k=0}^{n_o} c_i(p) c_k(q) [\text{imag } F(i\pi/L_e, -k\pi/L_o) - \text{imag } F(i\pi/L_e, k\pi/L_o)] \quad (134)$$

case 3:  $p$  odd and  $q$  even

$$m_{p,q} = -2j \sum_{i=0}^{n_o} \sum_{k=0}^{n_e} c_i(p) c_k(q) [\text{imag } F(i\pi/L_o, k\pi/L_e) + \text{imag } F(i\pi/L_o, -k\pi/L_e)] \quad (135)$$

case 4:  $p$  and  $q$  odd

$$m_{p,q} = 2 \sum_{i=1}^{n_o} \sum_{k=1}^{n_o} c_i(p) c_k(q) [\text{real } F(i\pi/L_o, k\pi/L_o) - \text{real } F(i\pi/L_o, -k\pi/L_o)] \quad (136)$$

In all four cases, approximately three quarters of the combined coefficients  $c_i(p)c_k(q)$  are zero since  $c_i(p) = 0$  when  $i > 0$  and even and  $c_k(q) = 0$  when  $k > 0$  and even. In all cases,  $m_{p,q}$  is a real number. If  $p$  is even and  $q$  is odd or

$p$  is odd and  $q$  is even then  $c_i(p)c_k(q)$  is imaginary and  $jc_i(p)c_k(q)$  is real. Sample spacing in the  $(\omega_x, \omega_y)$  - plane for the four cases above is shown in Figure 25. The dashed half square is the first zero of the two-dimensional sinc function which results from the average light intensity of the image. The transform is to be sampled at the intersection of all solid lines including the  $\omega_x$  and  $\omega_y$ -axis. The total number of samples depends on  $n_e$  and  $n_o$ . Notice that real or imaginary parts of the transform for samples which have the same  $\omega_x$ - coordinate but opposite  $\omega_y$ - coordinates are added or subtracted before being multiplied by  $c_i(p)c_k(q)$ .

Two Fortran computer programs have been written to compute the  $c_i(p)$ ,  $c_k(q)$  coefficients as defined by equations (105), (106), 107) and (108). These programs are listed in Appendix C.

The final topic which must be considered is how  $n_e$  and  $n_o$  are chosen to achieve a required accuracy in the approximations for  $[x^p]$ ,  $[y^q]$ . To this end, two additional Fortran computer programs were written to analyze the error function

$$e_{p,n}(x) = [x^p] - g_p(x) \quad (137)$$

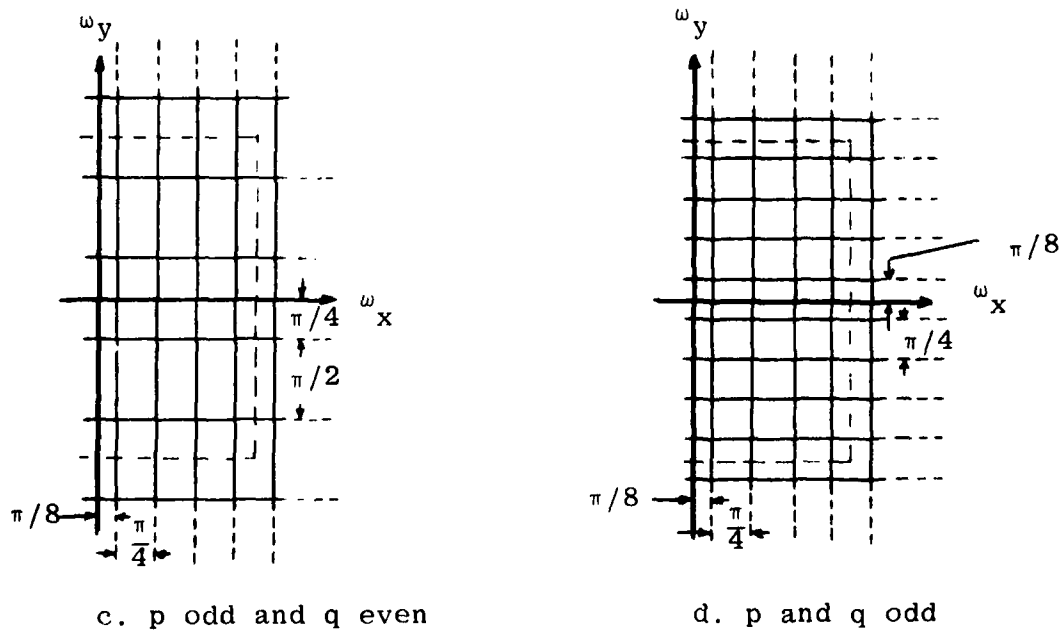
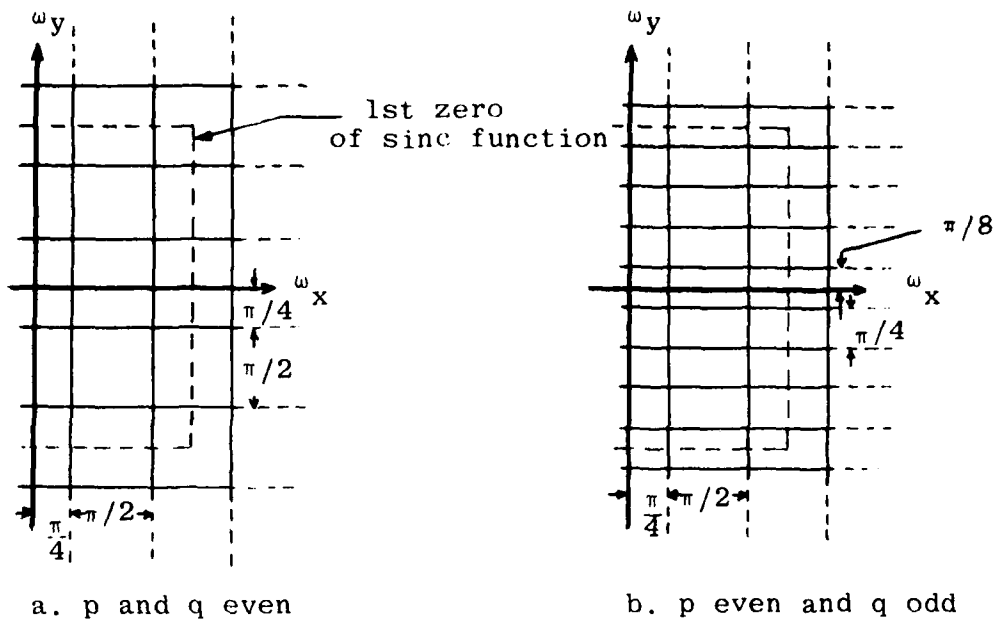
This is a function of  $x$ , the power  $p$  and the number of terms,  $2n+1$ , in the summation defining  $g_p(x)$ , equation (101). Now because of the form of  $g_p(x)$

$$e_{p,n}(x_i) = 0 \quad i = 0, \pm 1, \pm 2, \dots, \pm n; \quad x_i = iL/n \quad (138)$$

Because of the symmetries present it is easy to see that

$$|e_{p,n}(x)| = |e_{p,n}(-x)| \quad (139)$$

Hence, it is only necessary to examine  $e_{p,n}$  for positive values of  $x$ . If  $p$  is even then we are only interested in  $e_{p,n}$  over the interval  $0 \leq x \leq x_{n/2}$  which is right-half of the sensor aperture (normalized). If  $p$  is odd then we are, instead, interested in  $e_{p,n}$  over the interval  $0 \leq x \leq x_{n/4}$ . The Fortran



**Figure 25** - Fourier Transform Sample Spacing for Moment Computation

programs allow the user to enter  $p$ ,  $n/2$  or  $n/4$  and a mesh divisor  $m$ . The program then computes all the  $c_i(p)$  coefficients and evaluates  $e_{p,n}(x)$  at the points

$$x_i = iL/nm; i = 0, 1, \dots, nm/n_p \quad (140)$$

where  $n_p = 2$  if  $p$  is even and  $n_p = 4$  if  $p$  is odd. That is, the interval  $x_i \leq x \leq x_{i+1}$  is divided into  $m$  parts.

Using these Fortran programs a study was conducted to determine the effect of  $n$  on the maximum error observed at the  $1+nm/n_p$  grid points. The mesh divisor was fixed at  $m = 10$  for all cases. Then  $p$  and  $n$  were varied and the maximum error overall the grid points observed and tabulated. For the cases  $p = 1$  and  $n/4 = 1, 2$  and  $4$  the error function is shown in Figure 26. Notice that the maximum error decreases by at least a factor of two for every doubling of  $n$ . The tabulation of maximum error for the cases of  $1 \leq p \leq 8$  and  $n/n_p = 1, 2, 4$  and  $8$  is given in Table 3. From this table it is seen that the maximum error increases for  $n/n_p$  fixed and increasing  $p$ . Likewise, if  $p$  is held constant then the maximum error decreases with increasing  $n/n_p$ . Now  $n/n_p$  is the number of points in the interval  $0 \leq x \leq 1$  for which  $e_{p,n}(x) = 0$ . Notice that (approximately) the maximum error decreases by a factor of two when  $n/n_p$  is doubled and  $p$  held fixed. To investigate further, addition data was generated and tabulated in Table 4. From this table it is seen that the maximum error  $|e_{p,n}|$  will remain approximately the same if when  $p$  is doubled  $n/n_p$  is also doubled. (More accurately, the maximum error is increasing slightly as  $p$  and  $n/n_p$  are progressively doubled.) Also note that for (approximately) the same error,  $n/n_p$  should be the same for  $p$  even and  $p$  odd. Since  $n_p = 2$  for  $p$  even and  $n_p = 4$  for  $p$  odd this implies that

$$n_o = 2n_e \quad (141)$$

for (approximately) the same error. A maximum  $|e_{p,n}| = .03$  is an error of 3% of the full-scale value of  $|x^p|$  in the interval  $0 \leq x \leq 1$  which is 1.0.



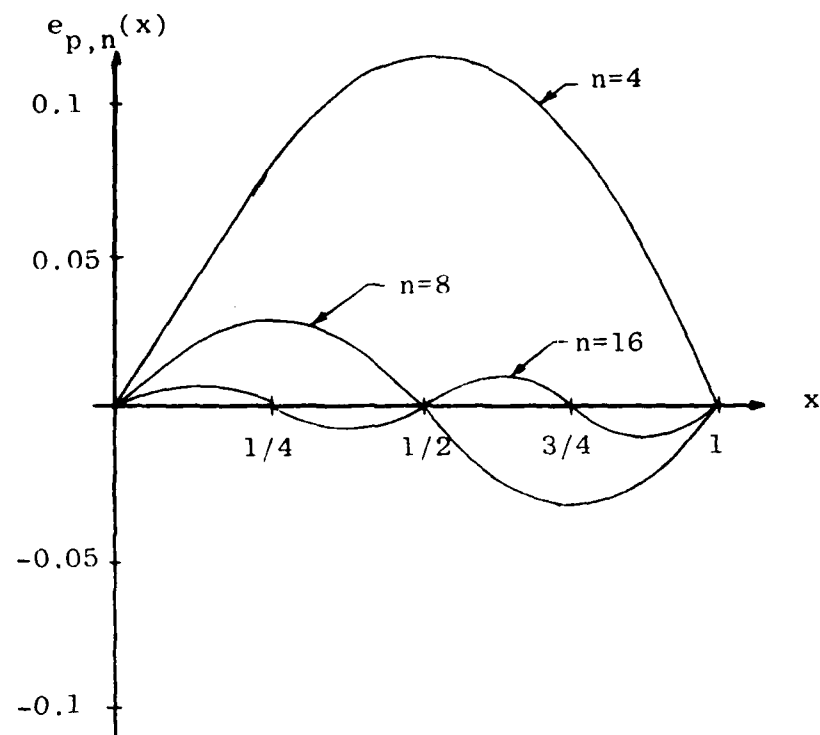


Figure 26 -  $e_{p,n}(x)$  for  $p = 1$  and  $n/4 = 1, 2$  and  $4$

$\begin{matrix} n/n_p \\ p \end{matrix}$	1	2	4	8
1	.1173166	.0353821	.0098962	.0026195
2	.0560095	.0172765	.0045921	.0011702
3	.2697230	.0526488	.0138003	.0035090
4	.3059095	.1007140	.0273291	.0070052
5	.4319005	.1697540	.0457383	.0116845
6	.4288390	.2163066	.0656872	.0173221
7	.5176162	.2795536	.0905453	.0241607
8	.5232109	.3136496	.1143713	.0317055

Table 3 Maximum  $|e_{p,n}(x)|$  vs.  $p$  and  $n/n_p$

Now refer to Figure 25 in order to relate  $n/n_p$ , and  $n$  to the spatial frequency bandwidth of the acousto-optic sensor. The first zero of the average intensity sinc function is the location of spatial frequencies of one line pair per aperture. The intersection of this zero with the  $\omega_x$  and  $\omega_y$ - axis is

$$\omega_x = \omega_y = \pi \quad (142)$$

A sensor with bandwidth of  $n_b$  line pairs per aperture has usable bandwidth

$$0 \leq \omega_x \leq n_b \pi$$

$$-n_b \pi \leq \omega_y \leq n_b \pi$$

Now  $n_e = 2n/n_p$  and  $n_o = 4n/n_p$ . The bandwidth required for  $g_p(x)$  is

$$p \text{ odd: } n_o \pi / 8 = (4n/n_p) \pi / 8 = (n/n_p) \pi / 2 \quad (143)$$

$$p \text{ even: } n_e \pi / 4 = (2n/n_p) \pi / 4 = (n/n_p) \pi / 2 \quad (144)$$

Equating with  $n_b \pi$  yields

$$n/n_p = 2n_b \quad (145)$$

Referring to Table 4, in order to compute up to 16th order moments with maximum  $|e_{p,n}| \leq .0336$  requires  $n/n_p = 16$ . Equation (145) then implies that the sensor bandwidth must be at least 32 line pairs per aperture along both the  $\omega_x$  and  $\omega_y$ -axis. Continuing with this example, if the sensor bandwidth was 16 line pairs per aperture then only 8th order moments could be computed with the same maximum  $|e_{p,n}|$ . However, if the error criteria was relaxed to maximum  $|e_{p,n}| \leq .0672$  then 16th order moments could be computed with a 16 line pair per aperture device. Table 3 and 4 or the computer programs of Appendix C can be used to determine the bandwidth required

p	n/n <sub>p</sub>	max  e <sub>p,n</sub>
2	2	.0172765
4	4	.0273291
8	8	.0317055
16	16	.0335946

a. p even

p	n/n <sub>p</sub>	max  e <sub>p,n</sub>
1	2	.0353821
3	4	.0138003
7	8	.0241607
15	16	.0296186

b. p odd

Table 4 - Maximum |e<sub>p,n</sub>(x)| vs. p = n/n<sub>p</sub> or p = n/n<sub>p</sub> - 1

given the highest order moments to be computed along with the largest allowable error. This analysis has assumed that the Fourier transform can be computed with perfect accuracy. Errors in the transform will lead to additional errors in the computed moments.

Since the two-dimensional moments are computed as weighted sums of two-dimensional Fourier transform components, the implementation of a preprocessor using an A-0 device can take the form of Figure 21. The A-0 device is used to compute Fourier components and the weighting and summation is carried out in the digital processor. The digital processor is also used to compute the moment invariants. An advantage of this method of computing moment invariants is that the same hardware configuration can be used to compute moment invariants as well as to compute the invariant signatures developed in Section III. 3.C. This would be desirable in applications where a decision is to be made based on prefilter outputs derived from more than one algorithm.

The algorithm for computing moments from the Fourier transform can be modified to provide some noise filtering. In aerial images it is often the case that the features of a signal that permit discrimination generally have significant high spatial frequency content in some frequency band<sup>17</sup>. The noise contributes primarily to spatial frequencies outside this band. It may be desirable to weight the Fourier transform components to emphasize signal and de-emphasize noise. The algorithm permits a weight function of the form

$$W(\omega_x, \omega_y) = W_x(\omega_x)W_y(\omega_y) \quad (146)$$

to be applied to the Fourier transform components prior to summation. These weights should be combined with the coefficients  $c_i(p)$ ,  $c_k(q)$ . The composite coefficients are

$$\tilde{c}_i(p) = W_x(i\pi/L)c_i(p) \quad (147)$$

$$\tilde{c}_k(q) = W_y(k\pi/L)c_k(q) \quad (148)$$

These multiplications can be precomputed and stored. Hence, noise reduction can be added to the algorithm without the need for any additional on-line computations. Examples of  $W_x$  functions are

$$W_x(\omega_x) = \begin{cases} 0; & \omega_x < \omega_c \\ 1; & \omega_x \geq \omega_c \end{cases} \quad (149)$$

which is a high-pass filter and

$$W_x(\omega_x) = \omega_x \quad (150)$$

which is a derivative filter. It should be noted however that two-dimensional filters such as

$$W_x W_y = \omega_x \omega_y \quad (151)$$

have a directional bias. This is not totally consistent with the desire to compute invariant signatures. This topic needs further investigation.

Next, an alternative approach to computing image moments will be considered. This approach utilizes the development based on the method of trigonometric interpolation which was derived above.

The A-0 devices which have been considered in this report which compute the function given by equation 1 are linear in  $g$  and  $h$  so that

$$\iint \hat{I}(x,y) \sum_{i=0}^n \sum_{k=0}^n g_i(t-x/v_x) h_i(t-y/v_y) dx dy = \sum_{i=0}^n \sum_{k=0}^n \iint \hat{I}(x,y) g_i(t-x/v_x) h_i(t-y/v_y) dx dy \quad (152)$$

This property allows an alternative means of computing moments. Rather than weighting and summing Fourier components in the frequency domain, the eigenfunctions can be weighted and summed in the time domain. The composite function is then applied to the A-0 device. The A-0 device output is then the

desired moment. The weighted eigenfunctions are given by equations (101) and (102). However, these are just the bandlimited approximations to  $[x^p]$  and  $[y^q]$  derived from the method of trigonometric interpolation. Hence, if the A-0 device electrical inputs are the approximations to  $[x^p]$  and  $[y^q]$  then the device output will approximate  $m_{p,q}$ . This statement needs two qualifications. First, the device output will only equal  $m_{p,q}$  at the time instant when the SAW or BAW are aligned as shown in Figure 24. That is, there is only one instant per period when the acoustic waves line up on the sensor to give the correct weighting function. The second qualification arises because of the image sampling caused by the metal grid pattern on the sensor. This was discussed in Section II. The origin of Fourier space is translated by  $f_{o,x}$  and  $f_{o,y}$ . To likewise translate the spectra of  $g_p(x)$  and  $h_q(y)$  they must be multiplied by  $e^{j2\pi f_{x,o}}$  and  $e^{j2\pi f_{y,o}}$  respectively. In an actual implementation, the complex exponentials would be replaced with real sinusoids and a synchronous detector used to preserve the phase information. Such a circuit is described in reference 18. Hence, the sensor drive signals should be of the form

$$g(t) = g_p(t) \sin 2\pi f_{x,o} t \quad (153)$$

$$h(t) = h_q(t) \sin 2\pi f_{y,o} t \quad (154)$$

Now since  $g_p$  and  $h_q$  have been designed to be bandlimited,  $g$  and  $h$  will also be bandlimited and  $n$  can be chosen to achieve maximum accuracy given the A-0 device bandwidth. Hence, all the previous analysis can be applied to this implementation also. The functions, equations (153) and (154) can be generated either by performing an analog multiplication of  $g_p$  or  $h_q$  with  $\sin 2\pi f t$  or by precomputing time samples of  $g(t)$  and  $h(t)$  and storing these samples in a fast digital memory. The memory would then be read, D/A converted and the analog samples smoothed and applied to the A-0 sensor. The sensor

output should be sampled in synchronism with the input so that this output is sampled and held at the instant when the correct portion of  $g_p(x)h_q(y)$  is on the active sensor area. The implementation of such a preprocessor is shown in Figure 27.

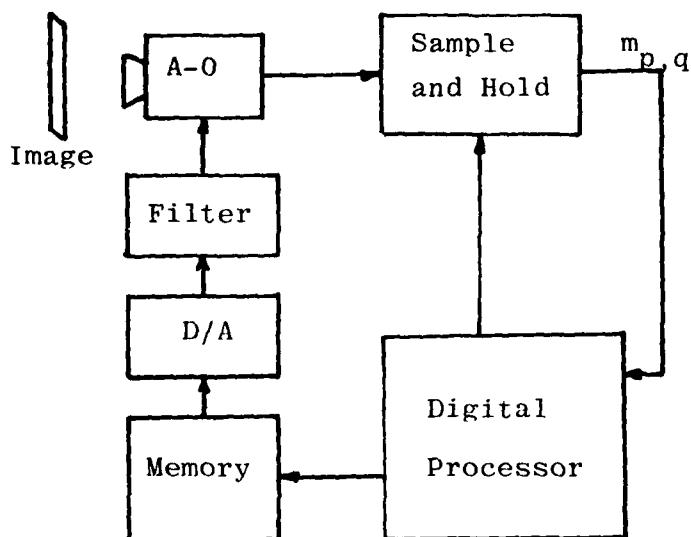


Figure 27 - Implementation: Method of Moments

#### D. Summary

This section began with a general discussion of prefilters. It was shown that the desirable property of feature isolation could be achieved by breaking the image into a number of smaller, overlapping views. By this means signal-to-noise ratio can be improved. Since A-0 sensors cannot match the accuracy of a digital processor, feature isolation is useful since it relaxes the noise filtering requirements of the prefilter algorithm. However, the price paid is the need to process a number of views rather than a single aerial image.

The concept of feature invariance was introduced. A prefilter which is invariant to changes in the feature such



as translation or rotation or scale is very desirable. This is because the number of distinct signals at the prefilter output is greatly reduced which simplifies the decision processor. However, invariance usually leads to an undesirable reduction in separation between feature classes.

Because of the function performed by A-0 sensors they are best suited to implement prefilters which utilize a separable transformation of the image of the form of equation (1). A number of such transformations were investigated.

The Hadamard transform is characterized as an orthonormal transform for which all the elements of the  $n \times n$  transformation matrix  $H_n$  are either + or -1. Because of this property, it is well-suited for digital implementation since additions rather than multiplications are required. Since the A-0 sensors performs general multiplications as easily as multiplications by  $\pm 1$ , this special property of  $H_n$  is of no advantage in an A-0 implementation. Since A-0 devices are bandlimited whereas Walsh functions are not bandlimited, A-0 devices do not appear to be the most optimum means for computing Hadamard transforms.

A classical means for feature extraction is the matched filter. However, the A-0 implementation of the matched filter is not desirable since both forward and inverse Fourier transforms are required. The matched filter for the image autocorrelation function leads to a much simpler A-0 implementation since only a forward transform is required. This version of the matched filter is invariant to feature translation but not rotation or scale.

More complex algorithms were developed which utilize Fourier transform samples to compute feature signatures which are invariant to translation, rotation and scale. Since the transform is sampled, the amount of information in the image is reduced to the point where a modest digital processor can compute the remaining steps in the algorithm. These algorithms are well suited for an A-0 device implementation consisting of

an A-0 sensor and a digital post processor.

Image two-dimensional moments were introduced. Certain combinations of these moments can be invariant to feature translation, rotation and scale change. Two methods of computing these moments from the Fourier transform were presented. The second method is more accurate for A-0 implementation since it utilizes Fourier samples over the entire bandwidth of the sensor. An alternative method of computing moments by time-weighting the electrical inputs to the A-0 sensor was developed. Using this method, moments can be computed potentially much faster than by the Fourier transform method. However, a more complex hardware configuration is required.

In summary, there appear to be a number of potentially useful feature extraction algorithms which can be effectively implemented using A-0 sensors.

#### IV. FEATURE EXTRACTION EXPERIMENTS WITH DEFT SENSORS

##### A. Introduction

For the purpose of sensor and algorithm evaluation, Deft Laboratories Inc. has developed a microprocessor-based Deft sensor operating system. This system consists of a MC6800-based microprocessor system, two digitally controllable sine wave generators, a Deft sensor and an electronics module for signal filtering and amplification. The system can interface with a tape recorder, teletype, storage CRT and an X-Y plotter. Assembly language programs can be written to implement signal processing algorithms. In addition, some resident software is available to make measurements of the Deft sensor output and to make pseudo-three dimensional plots of the magnitude of the Fourier transform.

Since this facility is available, it was decided to program one or more of the prefilter algorithms under study in order to make a preliminary evaluation of the concept of implementation developed in Section III. The present experimental set-up has two limitations. First, the system is limited to evaluating Deft sensors. Second, the sensor drive functions  $g(t)$  and  $h(t)$  are limited to sinusoids. Because of this second limitation experiments were restricted to those algorithms which characterize the image by its two-dimensional Fourier transform. Algorithms which were programmed are the method of invariant Fourier signatures and the method of invariant moment signatures. However, because of a shortcoming of the Deft sensors which were available during this study, it was not possible to compute image moments with any accuracy. This shortcoming is presently being eliminated in a new sensor design. More will be said about present shortcomings in the next section. The method of invariant Fourier signatures is less sensitive to this problem so that a series of experiments were conducted. The results of these experiments are detailed here.

First, however, the computer programs which were written will be described. This will be followed by a description of the experiments which were conducted and the results of these experiments. The section closes with some conclusions.

## B. Feature Extraction Experimental Computer Programs

### 1. Method of Invariant Moment Signatures

The flow diagram of this program is rather straightforward and is shown in Figure 28. When the program is entered it requests that the sensor view a uniform image for calibration. When this image is in place the user types "C" and the program makes a series of measurements near the transform origin for the purpose of removing a linear phase term which is present in the Deft sensor output. This phase function is approximately of the form

$$\phi(\omega_x, \omega_y) = k_x \omega_x + k_y \omega_y + k_0 \quad (155)$$

The computed Fourier transform is

$$\hat{F}(\omega_x, \omega_y) = F(\omega_x, \omega_y) e^{j\phi(\omega_x, \omega_y)} \quad (156)$$

where  $F$  is the desired Fourier transform.

This phase function is undesirable. The constant  $k_0$  is measured at the Fourier origin and removed from all data points. The linear phase also must be measured and its effects removed from each data point. The method of trigonometric interpolation is implemented in this program. Refer to Figure 24. If linear phase is not removed then the effect is to shift the aperture of the sensor out of the normalized aperture centered at  $x = y = 0$  and extending to  $x = \pm 1, y = \pm 1$ . The image will then, in effect, be multiplied by  $[x^p]$ ,  $[y^q]$  outside the interval where these functions equal  $x^p, y^q$ . The result will no longer be the image moment. To remove the linear phase,  $k_x$  and  $k_y$  are measured by sampling the transform at two points on the  $\omega_x$ -axis and two points on the  $\omega_y$ -axis near the Fourier origin where the transform has a large magnitude. Since a

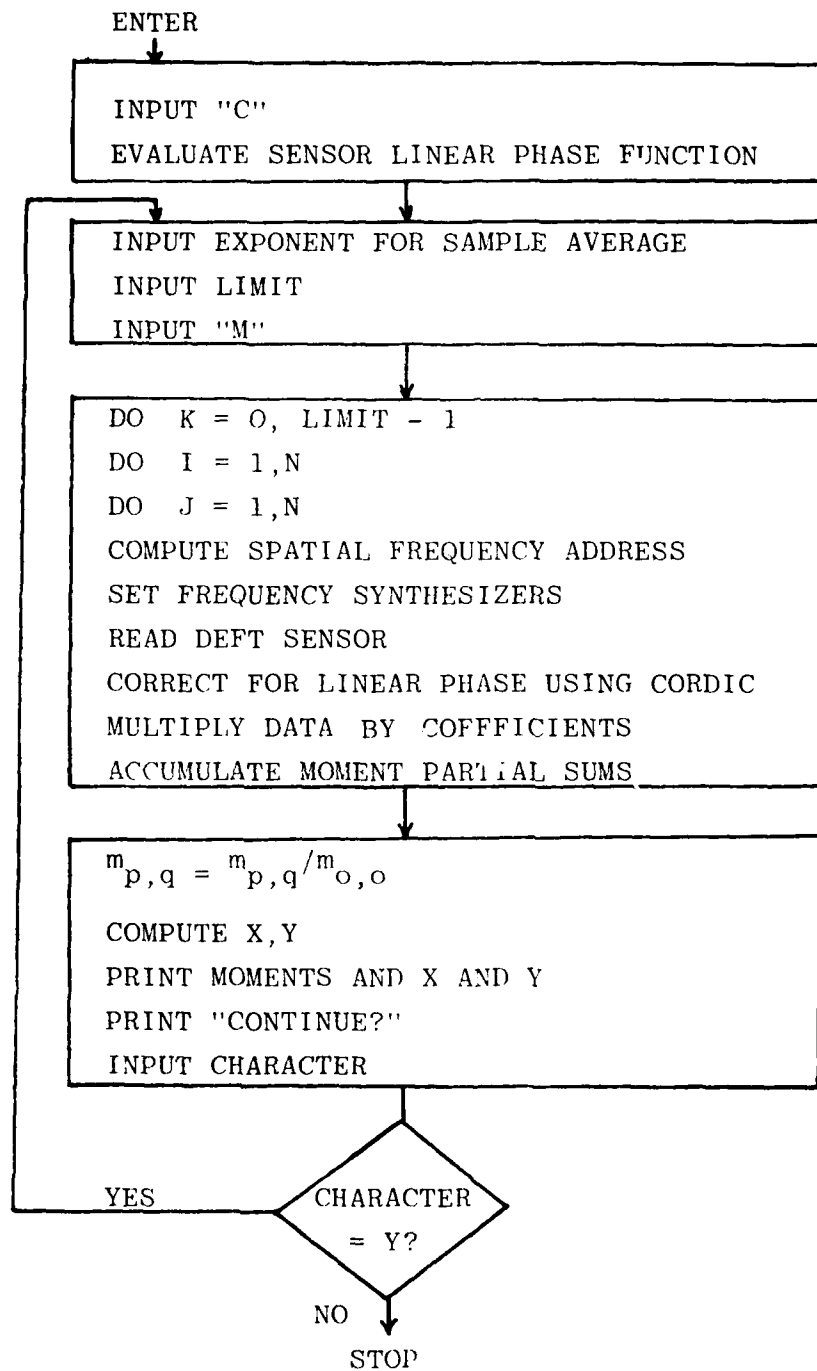


Figure 28 - Flow Diagram: Invariant Moment Signatures

uniform image has a transform which is a real function, any linear phase measured is due to the sensor and its associated electronics and should be removed. For example, if the sensor is measured at  $(0,0)$ ,  $(\tilde{\omega}_x, 0)$  and  $(0, \tilde{\omega}_y)$  with result

$$\hat{F}(0,0) = A(0,0)e^{j\phi(0,0)} \quad (157)$$

$$\hat{F}(\tilde{\omega}_x, 0) = A(\tilde{\omega}_x, 0)e^{j\phi(\tilde{\omega}_x, 0)} \quad (158)$$

$$\hat{F}(0, \tilde{\omega}_y) = A(0, \tilde{\omega}_y)e^{j\phi(0, \tilde{\omega}_y)} \quad (159)$$

then

$$k_o = \phi(0,0) \quad (160)$$

$$k_x = \frac{\phi(\tilde{\omega}_x, 0) - \phi(0,0)}{\tilde{\omega}_x} \quad (161)$$

$$k_y = \frac{\phi(0, \tilde{\omega}_y) - \phi(0,0)}{\tilde{\omega}_y} \quad (162)$$

The program automatically makes these measurements and stores the corrections  $k_o$ ,  $k_x$ , and  $k_y$ . The computed linear phase is then removed from each transform sample during the course of the measurements when the test image is in place. The samples which are available to the microprocessor from the sensor are in the form of the real and imaginary components of the transform. To remove linear phase, these samples must be converted to a magnitude and phase representation, the phase correction computed and applied and then the result converted back to real and imaginary format. The conversion is accomplished using the CORDIC algorithm. The total time involved per correction is about the same as the time required to compute two multiplies.

Once the parameters  $k_o$ ,  $k_x$  and  $k_y$  have been measured the program request a sample average which is a power of two ( $2^{**S}$ ). The user then inputs the exponent. Subsequent transform values will be sampled  $2^{**S}$  times per spatial frequency and averaged to improve the signal-to-noise ratio.

AD-A110 322

DEFT LABS INC EAST SYRACUSE NY

F/S 20/1

ACOUSTO-OPTIC TECHNOLOGY FOR TOPOGRAPHIC FEATURE EXTRACTION AND--ETC(U)

MAR 81 A L MOYER

DAAK70-79-C-0160

UNCLASSIFIED

0102-A002

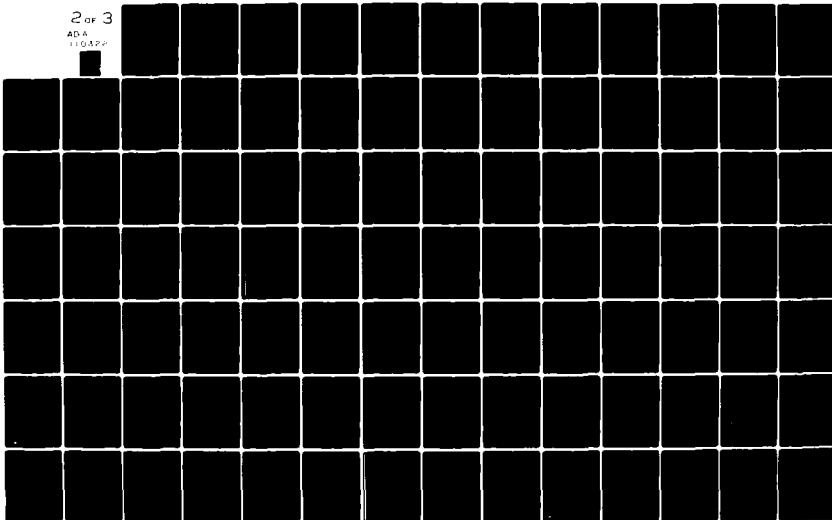
ETL-0256

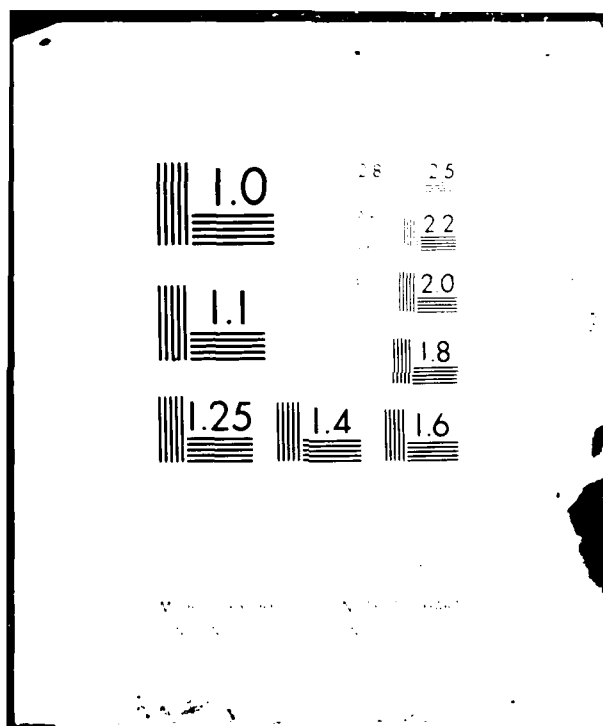
NL

2 of 3

AD A

110422







The program in its present form is capable of computing all image moments from  $m_{0,0}$  to  $m_{8,8}$ . To save time, if not all these are required, the user can specify a limit  $\leq 8$ . Then only moments from  $m_{0,0}$  to  $m_{\text{limit}-1, \text{limit}-1}$  will be computed.

Next the program requests the user to position the test image and type "M". The program then computes the two-dimensional moments using equations (133), (134), (135) and (136). These moments are then normalized by  $m_{0,0}$ . It is easy to see that

$$m_{0,0} \geq m_{p,q} \quad (163)$$

for any p and q. Hence,

$$\frac{m_{p,q}}{m_{0,0}} \leq 1 \quad (164)$$

Then two moment invariants are computed. These are

$$X = a^2 + b^2 \quad (165)$$

$$Y = a^2 - b^2 \quad (166)$$

where  $a^2 + b^2$  are defined by equations (82) and (83). The computed moments and X and Y are then printed out. The program then asks the user if he would like to continue. If he types a "Y" the program loops back and asks for a new sample average. If any other character is typed the program jumps to the monitor program and displays "\*". A sample run of the program is shown in Figure 29. The notation  $2^{**\pm(\text{integer})}$  means that the fixed point number to the left is to be multiplied by  $2^{**\pm(\text{integer})}$ . This is floating point using powers of two rather than ten. In the figure the case LIMIT = 3 is shown. If this parameter were larger then a larger table (up to 8 x 8) would be printed out.

POSITION UNIFORM IMAGE. THEN TYPE C C  
COMPUTES M(P,Q) MOMENTS FROM TRANSFORM

SAMPLE AVERAGE IS 2\*\*3

LIMIT = 3

POSITION TEST IMAGE. THEN TYPE M M

M(P,Q)

	0	1	2	3	4	5	6	7
0	+.9999	+.0055	+.1848					
1	-.1204	-.0181	-.0707					
2	+.1851	-.0468	+.0183					

X = +.7109 2\*\*-01 Y = +.6048 2\*\*-04

CONTINUE? Y

Figure 29 - Moments for Feature Extraction:  
Sample Run

For the particular implementation of the algorithm given in the program listed in Appendix B, the parameters  $n_e$  and  $n_o$  were set to be  $n_e = 20$ ,  $n_o = 40$  independent of  $p$  or  $q$ . This simplifies the computer program but results in run times which are longer than necessary because lower order moments can be accurately computed with much smaller values for  $n_e$ ,  $n_o$ . The coefficients which were used in the program are listed in Tables 5 and 6. These coefficients can be utilized by a Deft sensor with bandwidth of 10 line pair along the  $\omega_x$ -axis and  $\pm 10$  line pairs along the  $\omega_y$ -axis. (See Section III. C.4 for a discussion.) Since the sensor which was available had a resolution of at least 20 line pairs along the  $\omega_x$ -axis and  $\pm 10$  line pairs along the  $\omega_y$ -axis, the available bandwidth was sufficient for this application. Since  $n_e$  and  $n_o$  were too large for the lower power  $p$  and  $q$ , most of the coefficients in these cases are very small except for a few which are associated with transform samples near the Fourier origin. For example, consider  $c_k(2)$ . In this case only the first two or three coefficients are significant. Hence, lower order moments are insensitive to high spatial frequencies, as would

k	$c_k(0)$	$c_k(2)$	$c_k(4)$	$c_k(6)$
0	1.0000000	1.0000000	1.0000000	1.0000000
1	0.0	-0.5160232	-0.5864970	-0.6093051
3	0.0	0.0191080	0.1043204	0.1456567
5	0.0	-0.0041213	-0.0239249	-0.0501646
7	0.0	0.0014943	0.0088168	0.0202132
9	0.0	-0.0006940	-0.0041216	-0.0097799
11	0.0	0.0003693	0.0022002	0.0053087
13	0.0	-0.0002108	-0.0012579	-0.0030627
15	0.0	0.0001213	0.0007249	0.0017742
17	0.0	-0.0000636	-0.0003796	-0.0009316
19	0.0	0.0000198	0.0001184	0.0002910

Table 5 - Program Coefficients for p,q Even

k	$c_k(1)$	$c_k(3)$	$c_k(5)$	$c_k(7)$
1	-0.8109863	-0.8729509	-0.8872015	-0.8926364
3	0.0904811	0.2270004	0.2633326	0.2780638
5	-0.0328427	0.0826661	0.1262836	0.1463572
7	0.0169653	-0.0288055	-0.0660045	-0.0868873
9	-0.0104343	-0.0107975	-0.0356715	-0.0539768
11	0.0071316	0.0060242	0.0202588	0.0345133
13	-0.0052356	0.0042429	0.0124037	0.0227422
15	0.0040498	-0.0027277	-0.0081753	-0.0155041
17	-0.0032614	-0.0016602	-0.0056615	-0.0109329
19	0.0027129	0.0011839	0.0040415	0.0079358
21	-0.0023181	0.0009504	0.0029583	0.0058911
23	0.0020269	-0.0007069	-0.0022143	-0.0044456
25	-0.0018081	-0.0004914	-0.0016811	-0.0033913
27	0.0016420	0.0003742	0.0012820	0.0025989
29	-0.0015157	0.0003088	0.0009746	0.0019850
31	0.0014201	-0.0002323	-0.0007317	-0.0014942
33	-0.0013495	-0.0001532	-0.0005325	-0.0010885
35	0.0012995	0.0001024	0.0003619	0.0007408
37	-0.0012675	0.0000703	0.0002099	0.0004306
39	0.0012520	-0.0000271	-0.0000687	-0.0001412

Table 6 - Program Coefficients for p,q Odd

be expected. The coefficients in Tables 5 and 6 are stored in the computer program in Tables with labels C01, C23, C45 and C67. For example, C45 holds the coefficients for  $c_k(4)$  and  $c_k(5)$ .

## 2. Method of Invariant Fourier Signatures

Since this algorithm uses the magnitude (or magnitude squared) of the Fourier transform, the phase correction described in the preceeding section is not required.

The overall flow diagram for the computer program is shown in Figure 30. A detailed flow diagram of projection computation is shown in Figure 31. The program contains a number of parameters which can be set to control the algorithm used. Projections can be computed which are the result of integration of the Fourier transform magnitude along radial lines, around (semi) circles or along spirals.

When the program is entered, the user is asked to specify a sample average as a power of two, S. Then all Fourier samples will be taken  $2^{**}S$  times and averages to improve signal-to-noise ratio. The user is then asked to specify circles or not circles. ("Circles" is a special case which requires different logic in the program.) If circles are chosen then the projection to be computed will contain  $2^{**}5$  samples because the program always uses 32 radius values which are precomputed and stored in Table RVECT. The values of these radii satisfy the relationship given in equations (52) and (53). The projection results from integrating around the  $2^{**}5$  semicircles.

If circles are not chosen the user can choose the parameter M in which case the projection will consist of  $2^{**}M$  samples. M should not be chosen larger than 6. M=5 is a good value to adequately sample the Fourier transform. If M=5 running time of the program is under one minute per image. The projection results from integrating the transform magnitude along  $2^{**}M$  radial lines or spirals.

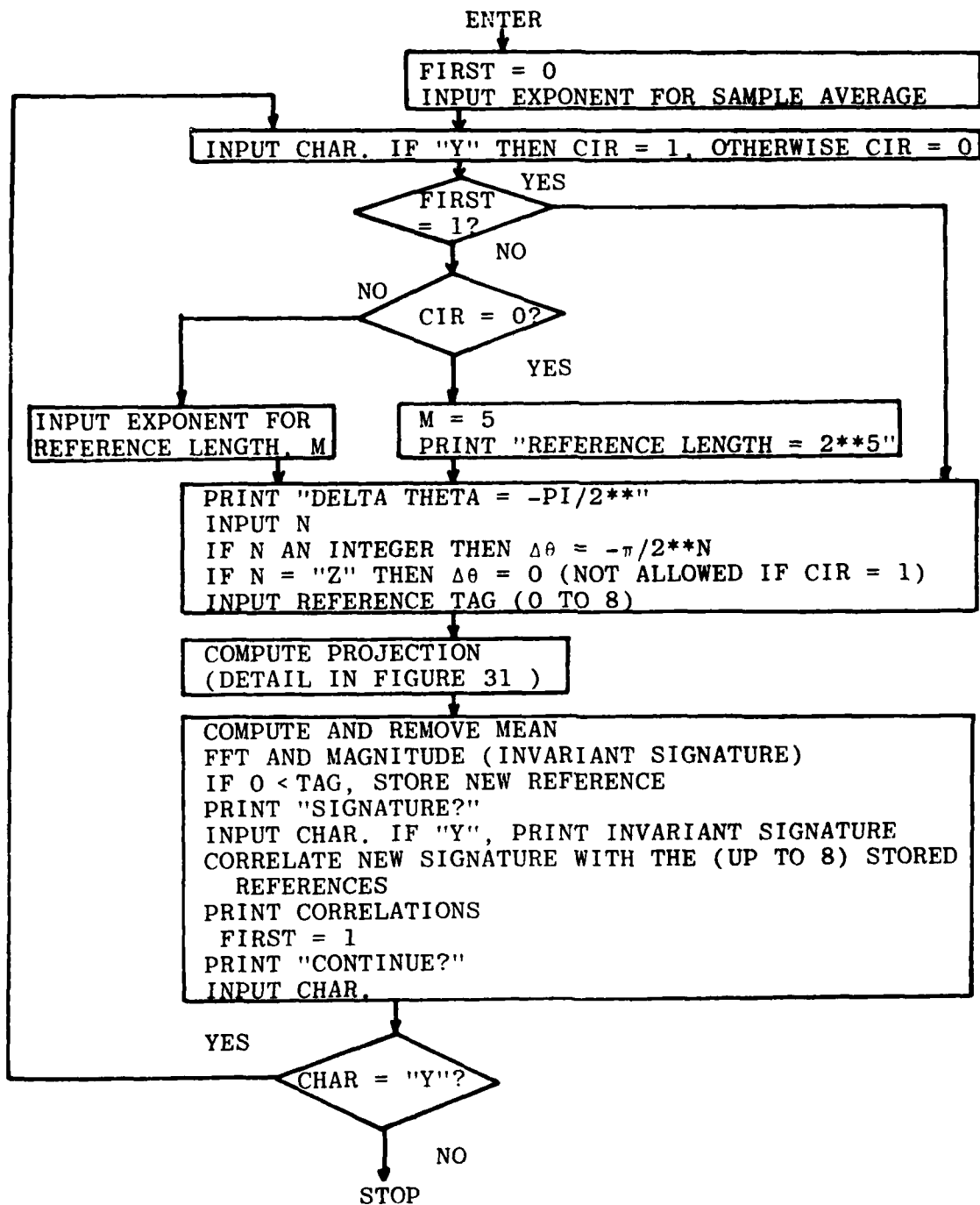


Figure 30 - Flow Diagram: Invariant Fourier Signatures

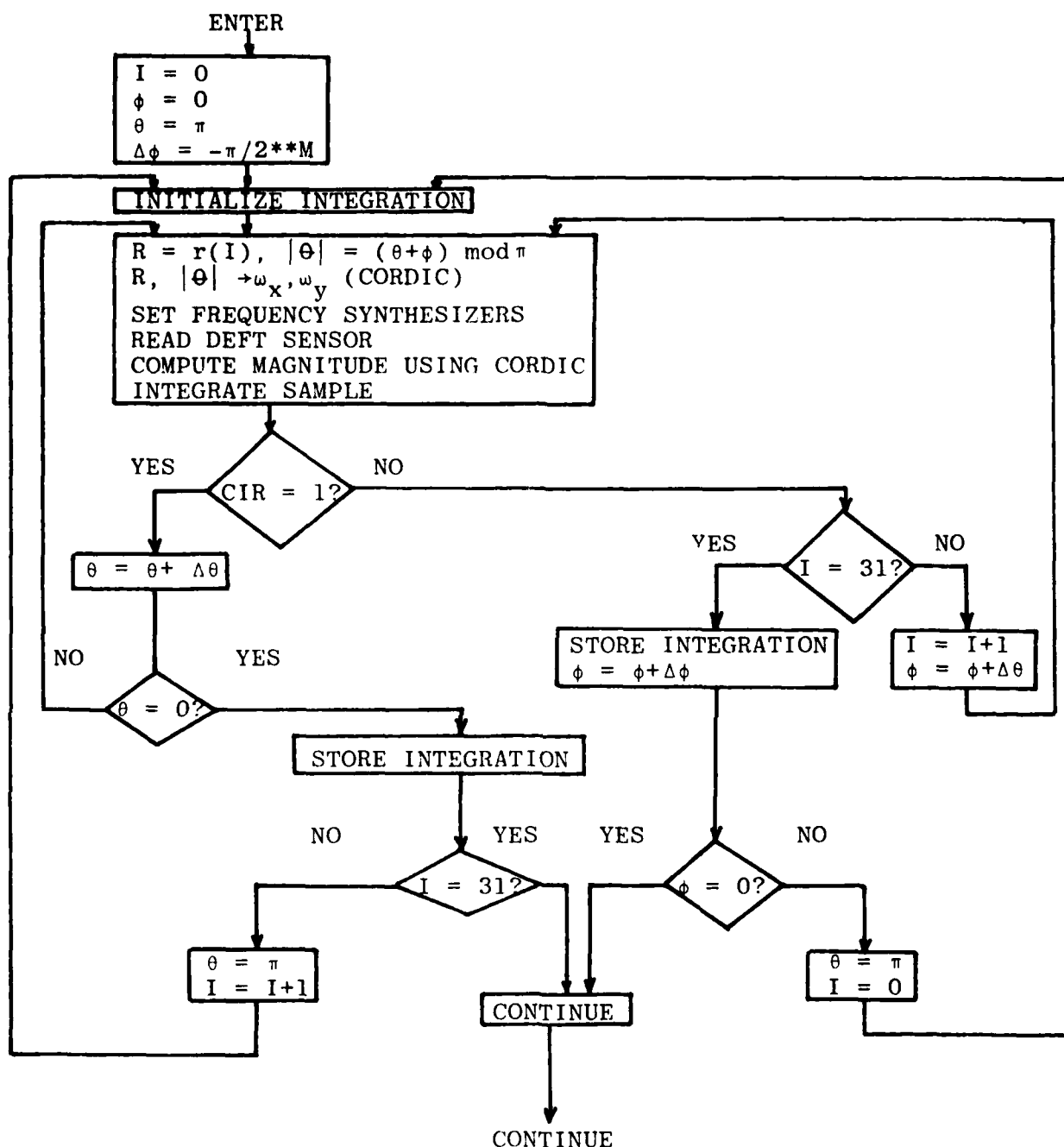


Figure 31 - Flow Diagram Detail: Projection Computation

The program then asks for the parameter  $\Delta\theta$ .  $\Delta\theta$  has two functions depending on whether circles are chosen or not chosen. Consider "circles" first. In that case  $\Delta\theta$  gives the angular spacing between samples along each semicircle. It also determines the number of samples which will be integrated along each semicircle. The finer the angle, the more the number of samples and the program running time. Again,  $\Delta\theta = -\pi/2^{**}5$  gives good sampling of the transform with a run time of under one minute. The polar coordinate of the kth sample along the i-th semicircle is given by

$$(r, \tilde{\theta}) = (r_i, (\pi + k\Delta\theta) \bmod \pi) \quad (167)$$

where  $r_i$  is the i-th radius. Angles are mod  $\pi$  because the transform is symmetric across the origin.

Now consider the case of "not circles". In that case  $\Delta\theta$  gives the amount of skew for the spirals. Each spiral consists of 32 samples (one for each of the 32 radii.) The polar coordinates of the i-th sample along the k-th spiral is given by

$$(r, \tilde{\theta}) = (r_i, (\pi + i\Delta\theta + k\Delta\phi) \bmod \pi) \quad (168)$$

where

$$\Delta\theta = -\pi/2^{**}M \quad (169)$$

If  $\Delta\theta = 0$  then the contours of integration will be radial lines rather than spirals. To get  $\Delta\theta = 0$ , input "Z" when the program prints "DELTA THETA = -PI/2\*\*". To get  $\Delta\theta \neq 0$  enter the exponent instead.  $\Delta\theta = 0$  will not be accepted by the program when "circles" since this would result in an infinite loop.

Next the program requests a reference tag. The user responds by entering an integer from 0 to 8. This is stored in variable TAG. If TAG > 0 then after the invariant signature is computed it will be stored in a table in position TAG. Up to 8 references can be stored in the table. References can be replaced at any time by a new reference. The purpose of



these references is to provide a means of performing some simple feature recognition experiments. The user can set up a number of reference images consisting of distinct features. The program then computes the invariant signatures and stores them in the table. Then the user can use the system to view new images which might be translated, rotated or scaled versions of the references. In that case the user enters 0 for the reference tag. If TAG=0 the signature is then correlated against all the reference signatures. Let  $\{r_i\}$  be the  $2 \times M$  samples of a reference and  $\{x_i\}$  be the samples of a new signature. Then the correlation is defined by

$$\text{correlation} = \frac{\sum_{i=1}^{2^m} r_i x_i}{\sqrt{\sum_{i=1}^{2^m} r_i^2} \sqrt{\sum_{i=1}^{2^m} x_i^2}} \quad (170)$$

This correlation is the basis for a simple decision processor. The signature is correlated with all reference signatures. These correlation coefficients are printed out by the program. The user can then observe these values and decide if the new image is a translated, rotated or scaled version of one of the references. A simple decision rule is to look for the largest correlation. If it is bigger than some threshold then the test image is considered to contain the feature corresponding to the reference signature at which maximum correlation occurred.

Returning to the discussion of the program, once the user enters the reference tag the program then computes the invariant signature. Sampling of the transform and integration of samples occurs in that portion of the program shown in Figure 31. There are two nested loops, one for each Fourier sample and one for each contour. The transform spatial frequency address is first computed in polar coordinates and then converted to rectangular coordinates using the CORDIC algorithm. The sensor is addressed and data read. The sensor output is converted to magnitude and phase using the CORDIC

algorithm. The magnitude is then integrated using the trapezoidal rule. At the end of each contour, the resulting integration is stored by being pushed onto the microprocessor stack. At the end of the last contour there will be  $2**M$  samples in the stack.

Next, the program computes the mean of these samples and subtracts the mean from each sample. This improves the accuracy of the FFT algorithm which is applied to these samples. The algorithm uses fixed point arithmetic with a block floating point scaling scheme. If the mean is not removed the FFT output will have a large component in the zero frequency bin. This leads to poor scaling for the remainder of the FFT bins which are, typically, much smaller in magnitude. Removing the mean corrects for this problem.

The FFT output samples are then converted to magnitude and phase representation using the CORDIC algorithm. The resulting  $2**M$ -vector of real numbers is the desired invariant signature.

The program then prints "SIGNATURE?". If the user enters "Y" the invariant signature is printed out. If instead "N" is entered, this is skipped. The program then stores the signature of TAG>0. It then correlates with all reference signatures and prints out the correlation coefficients.

The program then prints "CONTINUE?" Entering a "Y" causes the program to ask for new parameters for another run. Entering an "N" causes a jump to the monitor and a "\*" is displayed.

A sample run of the program is shown in Figure 32. The algorithm used in this run was  $2**5$  radial lines. There was no sample averaging. The first three correlation tables represent three references. The fourth table gives the correlation of a misaligned feature with the three references. In the case shown, correlation with reference 1 was largest. The test object was reference 1 scaled by a factor of 0.9 in size.

FEATURE RECOGNITION  
DEFT/PROJECTION/FFT

SAMPLE AVERAGE IS 2\*\*0  
CIRCLES?N  
REFERENCE LENGTH=2\*\*5  
DELTA THETA=PI/2\*\*Z  
REFERENCE TAG=3  
SIGNATURE? N

CORRELATION  
1) 0.0000 2) 0.0000 3) 0.9999 4) 0.0000  
5) 0.0000 6) 0.0000 7) 0.0000 8) 0.0000

CONTINUE? Y  
CIRCLES?N  
DELTA THETA=-PI/2\*\*Z  
REFERENCE TAG=2  
SIGNATURE? N

CORRELATION  
1) 0.0000 2) 0.9999 3) 0.7623 4) 0.0000  
5) 0.0000 6) 0.0000 7) 0.0000 8) 0.0000

CONTINUE? Y  
CIRCLES?N  
DELTA THETA=-PI/2\*\*Z  
REFERENCE TAG=1  
SIGNATURE? N

CORRELATION  
1) 0.9999 2) 0.8576 3) 0.8727 4) 0.0000  
5) 0.0000 6) 0.0000 7) 0.0000 8) 0.0000  
CONTINUE?

Y  
CIRCLES?N  
DELTA THETA=-PI/2\*\*Z  
REFERENCE TAG=0  
SIGNATURE? N

CORRELATION  
1) 0.9795 2) 0.9364 3) 0.8578 4) 0.0000  
5) 0.0000 6) 0.0000 7) 0.0000 8) 0.0000

Figure 32 - Invariant Fourier Signatures: Sample Run

### C. Feature Extraction Experiments

The two computer programs which are described in the previous section were written for the purpose of performing some preliminary feature extraction experiments utilizing the algorithms developed in Section III and the hardware system described in this section. It was not the intent of these experiments to show detection of real features in aerial imagery. Rather, the purpose was to verify feasibility of both the algorithms and the sensor technology to detect features from a small set of controlled text patterns. In this way, directions for further improvements in both algorithms and sensors could be determined.

It was originally intended to perform experiments using the method of invariant Fourier signatures (IFS) and the method of invariant moment signatures (IMS). However, the program which computes moments from the Fourier transform gave poor results. The reasons for this were identified. It was determined that the program computes moments with large errors because of two shortcomings of the Deft sensors which were used in the experiment. Since being identified, steps are being taken to correct this problem in future Deft sensors. More will be said about this in Section V. Briefly, however, the two sensor shortcomings were the following:

1. Because of the current collecting metal bars used in the present sensor design strong reflections will occur in the surface acoustic wave propagating orthogonal to these bars. (Refer to Figure 2.) This is because the periodic structure

of these bars reinforces the small reflections which occur at each bar. The effect of these reflections on the Fourier transform is to multiply the transform by a weight function of, say  $W_x$ . That is, the function which is available is  $\tilde{F}$  where

$$\tilde{F}(\omega_x, \omega_y) = W(\omega_x) F(\omega_x, \omega_y) \quad (171)$$

Visually,  $|\tilde{F}|$  is a rippled version of  $|F|$ . This can be seen in any of the transform plots which are given in Section V. One problem caused by these ripples is that it is not possible to accurately determine the origin of the Fourier transform. The origin will always be located at the peak of the zero frequency sinc function. However, because of the ripples on this peak, the true maximum cannot be determined. It can be approximately determined by plotting the peak and visually determining the center.

The second problem with ripples is that the transform will contain errors because of  $W(\omega_x)$ . Refer to Tables 5 and 6. Notice that for the lower order moments the coefficients  $C_k$  are largest for small indices  $k$ . This implies that components of the transform nearest the origin will be most heavily weighted in computing these moments. Refer, for example to  $C_k(2)$ . It can be seen that the computed moment will depend primarily on the value of the transform at the origin minus the weighted value of the transform at  $\omega = \pi/4$  (See Figure 25) If the peak of a ripple occurs at  $\omega = 0$  and the trough of the ripple occurs at  $\omega = \pi/4$  then the computed moment will be larger than the correct value. Other situations leading to other types of errors can be envisioned.

By examining the transform plots in Section V it is obvious that the large ripples shown there would lead to large errors in computing image moments. This proved to be the case in attempts to perform experiments. For example, in some cases moments  $m_{p,q}$  where both  $p$  and  $q$  were even were computed to have negative values. However, these moments must always be positive.

The solution to this problem is to redesign the Deft sensor to eliminate the reflections. This is presently being done as will be detailed in the next section.

2. The second shortcoming arises from other inaccuracies in computing the Fourier transform. These will be discussed in Section V. As mentioned, when viewing a completely uniform image, the Deft sensor output contains a linear phase term which must be removed to compute moments. However, because the transform is computed with some error, the assumption of linear phase is only an approximation. In attempting to use the computer program it was determined that the assumption of linear phase was only approximately true. Hence, the undesirable phase function can only be approximately corrected for. This is not a serious problem in the computation of low-order moments since they depend primarily on transform samples near the Fourier origin where an accurate phase correction can be made. However, the accurate computation of moments above the first few would require much better phase corrections. It is difficult to presently make such corrections for two reasons. First, the uniform, white image used as a test pattern has a transform whose magnitude is large over only a limited region of Fourier

space. It is only possible to measure phase in this region with this test pattern. Accurate phase measurement over the entire transform would require a large number of accurately positioned test patterns. Secondly, the Fourier transform output of the sensor is only an approximately linear function of light intensity. Hence, phase corrections measured at one intensity may be in error when the image to be analyzed has another average intensity.

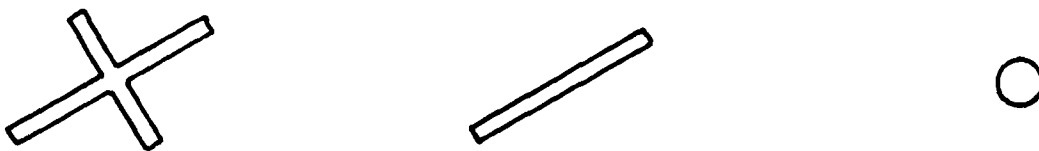
Efforts are underway to improve the accuracy of the computed transform. This should lead to more predictable and accurate phase measurement and correction. These efforts are discussed in the next section.

An alternative solution would be to compute the moments of the image autocorrelation function rather than of the image. That is, use the magnitude of the Fourier transform rather than the real and imaginary parts. This solution circumvents the phase problem but may be undesirable in that some image information is lost.

The remainder of this section is a discussion of the experiments conducted using the method of invariant Fourier signatures. In Part B.2 of this section the flexibility of the computer program was described. The algorithm developed in Section III can be configured in a number of ways using constants entered into the program. These constants control the sample averaging of the data, the angle  $\phi$  of the projection  $P_\phi$  and the number projections computed. In application, these parameters can be chosen to optimize the invariant signatures which are computed given a class of features to be detected. However, because of the scope of this program, only a limited

number of algorithm configurations were utilized. To be specific, two forms of the algorithm were used in all tests. These are integration along 32 radial lines and integration around 32 circles. No sample averaging was used in the experiments after it was determined that averaging did not affect the results. This was desirable since averaging increases program run time.

In all experiments, three reference objects were used. These references are shown in Figure 32. The images consisted of a dark background with white features.



a. "Crossroads"

b. "Road"

c. "Storage Tank"

Figure 33 Test Reference Features

These features are simple geometric objects. However, they resemble some important realistic features and were chosen on this basis. These features are a crossroads, a road and a compact, round structure such as a storage tank. The reference objects are meant to only be idealizations of these features. In particular, the images contain no background noise. A Viewlex projector with a rotating barrel was used to project the images which were in the form of slides. The 60 degree tilt of the images is a result of the Viewlex. The rotating barrel had a stop at which the slides were tilted 60 degrees. Rotation of images was then measured with respect to this stop. This 60



degree tilt has no significance.

The experimental set-ups which were used shown in Figure 33.

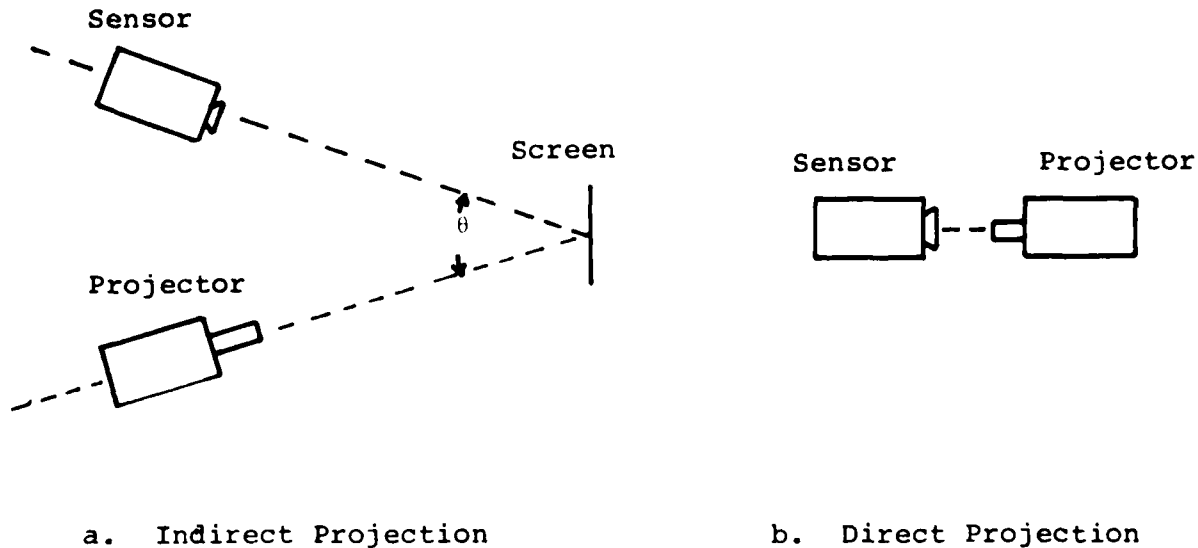


Figure 34 Experimental Set-Ups

In some experiments the image was projected on a screen and the sensor viewed the projected image. The angle  $\theta$  between projector and sensor was minimized to prevent image distortion. This set-up was used in the experiments where image scale was to be varied. This was accomplished by moving the projector along the dashed line and refocusing. The size of the image was measured on the screen using a ruler. In the other experiments the lens on the Deft sensor module was removed and the projector was placed in line with the sensor. The image was then focused directly, at close range, onto the Deft sensor. To do this it was necessary to stop down the projector lens using the  $f$ -stop adjustment. The module was mounted on a triangular rail so that it could be accurately translated at right angles to the dashed line in Figure

33b. This was used in the experiments where the feature to be detected was a translated version of the reference. The lens barrel was rotated in experiments where the feature to be detected was a rotated version of the reference.

During the period of time when experiments were conducted, two Deft sensors were available for test. These will be labeled sensor #1 and #2. The experiments will now be described. In all experiments the first step was to compute the three reference signatures using the images of Figure 32 in their reference positions. Then the system was presented with the same images but either rotated, scaled or translated or a combination of these. The signature was then computed and correlated against the three reference signatures. The reference at which the highest correlation occurred was then considered to be the detected feature. If this was the correct feature then the system "passed". If not, the system "failed". The specific experiments conducted were the following:

#### Image Rotation

Reference signatures were computed. The system was then presented with the same images but with the projector barrel rotated from 0 to 25 degrees in 5 degree steps.

#### Image Translation

Reference signatures were computed. The system was then presented with the same images but the sensor module was translated from 0 to 8 mm in 1 mm steps. The width of the active sensor area is 12.7 mm. Generally, translations of greater than 8 mm resulted in a significant part of some of the images falling outside the sensor active area.

### Image Scale

Reference signatures were computed. The projector was then moved closer to the screen and refocused so that a dimension on one of the references was reduced by a factor of  $k$ . This was repeated for  $k = \{.9, .8, .7, .6, .5\}$ . The system was presented with the same three images at each position of the projector.

### Image Translation with 5 Degree Offset

Reference signatures were computed. The projector lens barrel was then rotated 5 degrees. The system was then presented with the same images but with the sensor module translated from 0 to 8 mm in 1 mm steps.

### Image Translation with 10 Degree Offset

Same as above but with a 10 degree offset after the reference signatures have been computed.

### Image Translation with 15 Degree Offset

Same as above but with a 15 degree offset.

### Image Rotation with $k = .7$ Scale Offset

Reference signatures were computed. The projector was then moved toward the screen until all image dimensions were reduced by a factor of 0.7. The system was then presented with the same images but with the projector barrel rotated from 0 to 25 degrees in 5 degree steps.

The results of these experiments are given in Tables 7, 8, 9, and 10. These tables give the number of passes and fails as a function of sensor, algorithm, experiment and reference pattern. Both algorithms performed approximately equally.

$$\text{success rate} = \frac{\text{number of passes}}{\text{number of passes} + \text{number of failures}} \times 100\% \quad (172)$$

If only the experiments of rotation, translation and scale are considered then the success rate for all experiments and patterns are as follows:

Sensor #1, 32 radial lines: 81.7% success rate

Sensor #1, 32 circles: 81.7% success rate

Sensor #2, 32 radial lines: 86.7% success rate

Sensor #2, 32 circles: 80.0% success rate

As can be noted, both sensor performed approximately equally.

Experiment	Reference Pattern							
	1		2		3		Total	
	P	F	P	F	P	F	P	F
Rotation	4	2	5	1	6	0	15	3
Translation	7	2	9	0	6	3	22	5
Scale	5	0	5	0	2	3	12	3
Translation with 5 degree offset	2	7	9	0	8	1	19	8
Rotation with k=0.7 scale offset	4	2	6	0	4	2	14	4
Total: all experiments	22	13	34	1	26	9	82	23

Table 7 - Feature Extraction Experimental Results:  
Sensor #1, 32 Radial Lines

Experiment	Reference Pattern							
	1		2		3		Total	
	P	F	P	F	P	F	P	F
Rotation	6	0	6	0	6	0	18	0
Translation	7	2	6	3	7	2	20	7
Scale	3	2	5	0	3	2	11	4
Translation with 5 degree offset	5	4	6	3	8	1	19	8
Rotation with k=0.7 scale offset	6	0	5	1	3	3	14	4
Total: all experiments	27	8	28	7	27	8	82	23

Table 8 - Feature Extraction Experimental Results:  
Sensor #1, 32 Circles

Experiment	Reference Pattern							
	1		2		3		Total	
	P	F	P	F	P	F	P	F
Rotation	6	0	6	0	6	0	18	0
Translation	9	0	9	0	8	1	26	1
Scale	3	2	5	0	0	5	8	7
Total: all experiments	18	2	20	0	14	6	52	8

Table 9 - Feature Extraction Experimental Results:  
Sensor #2, 32 Radial Lines

Experiment	Reference Pattern							
	1		2		2		Total	
	P	F	P	F	P	F	P	F
Rotation	6	0	6	0	6	0	18	0
Translation	9	0	2	7	9	0	20	7
Scale	3	2	5	0	2	3	10	5
Total: all experiments	18	2	13	7	17	3	48	12

Table 10 - Feature Extraction Experimental Results:  
Sensor #2, 32 Circles

It was observed during the experiments that failure occurred more frequently where there was either a large translation, rotation or scale offset between references and test patterns. It has been determined that this is because of certain errors in the Deft sensor output. More will be said about these errors in Section V.

Some additional, more challenging experiments were performed using sensor #1. These experiments involved offsets of either translation and rotation or rotation and scale between reference and test images. The results of these experiments are also tabulated in Tables 7 and 8. For these experiments only, the success rates were the following:

Sensor #1, 32 radial lines: 73% success rate

Sensor #1, 32 circles: 73% success rate

as can be seen, the additional misalignment reduced the success rate somewhat. This is a consequence of the same Deft sensor errors alluded to above.

These experiments verify the premise of invariance to translation, rotation and scale. However, the success rates must be described as modest. Part of the difficulty can be traced to sensor performance, but another part may result from the algorithms which were used. One of the features of an ideal prefilter mentioned in Section III is that the prefilter increases the between-class separation of the feature classes. For the present experiments between-class separation is measured by the crosscorrelation of signatures of each of the reference patterns.

Table 11 Reference Pattern Correlations

		Reference Pattern		
		1	2	3
Reference Pattern	1	0.9754	----	----
	2	0.9225	0.9961	----
	3	0.8151	0.7508	0.9749

a. Sensor #1, 32 Radial lines

		Reference Pattern		
		1	2	3
Reference Pattern	1	0.9991	----	----
	2	0.9877	0.9989	----
	3	0.9495	0.9751	0.9970

b. Sensor #1, 32 Circles

		Reference Pattern		
		1	2	3
Reference Pattern	1	0.9943	----	----
	2	0.8551	0.9982	----
	3.	0.7819	0.9218	0.9696

c. Sensor #2, 32 Radial Lines

		Reference Pattern		
		1	2	3
Reference Pattern	1	0.9994	----	----
	2	0.9891	0.9996	----
	3.	0.9522	0.9827	0.9990

d. Sensor #2, 32 Circles



The crosscorrelations and auto correlations of the reference signatures are given in Table 11 for sensors #1 and #2 and for both algorithms. The entries represent the average of all values which were obtained in all the experiments tabulated in Tables 7 through 10. Notice that the crosscorrelation terms were not significantly smaller than the autocorrelation terms. This was particularly the case for the "circles" algorithm. Hence, between-class separation was not as large as might be desired. This drawback cannot be blamed on the sensor. Rather it is a consequence of the algorithms used. It is the feeling of this Author that the requirement of the second transform (FFT) tends to smooth the data in such a way that there are not large differences between the signatures of different reference patterns. It appears that by designing invariance into these algorithms and thereby decreasing the in-class separation, the between-class separation also decreases.

The experiments which were conducted used noise-free, idealized features. Since the performance reported here is only modestly successful it is evident that the present Deft sensor is not capable of detecting features in realistic aerial images making use of the algorithms developed in this report. It is possible that present sensors could be useful in more limited applications such as the detection of man-made vs. natural features. The utility of the Fourier transform in this application has been verified by Lendaris and Stanley.<sup>20</sup> Their algorithms do not require a second transform and exhibit greater between-class separation. However, they are not invariant to translation rotation and scale.

In order to take advantage of the more powerful algorithms developed in this report, a sensor which faithfully produces the Fourier transform is required. The basic requirements of the sensor output are the following:

1. The magnitude of the sensor transform should be invariant to feature translation.

2. If the feature rotates then the magnitude of the transform must also rotate.

3. In order to include filtering against additive noise the sensor must be linear. That is, if image  $I_1$  has transform  $F_1$  and image  $I_2$  has transform  $F_2$  then the combined image  $I_1+I_2$  has transform  $F_1+F_2$ .

## V. ACOUSTO-OPTIC SENSOR CAPABILITIES: PRESENT AND PROJECTED

### A. Introduction

This section deals with the present capabilities and limitations of the A-O devices considered in this study. Capabilities and limitations are considered in the context of feature extraction. Also considered are projected improvements and the probability of success.

Since the Deft sensor is presently receiving active development support by Deft Laboratories Inc., more detailed information can be provided on this sensor than the others which were discussed in Chapter 2. Recent and present Deft sensor development has been funded by NASA and by internal Deft Laboratories Inc. funds. The level of NASA funding for Deft sensor development in the last year is \$130,000.

This chapter begins with the elastobirefringent light valve. The Deft sensor is covered in the following section. The Thomson - CSF sensor can be thought of as a specialized Deft sensor so that it will not receive a separate discussion.

### B. Elastobirefringent Light Valve

This sensor is not presently receiving development support. Devices which have been developed to date should be considered experimental. In order to apply this sensor to feature extraction present device limitations must be overcome and manufacturing procedures developed. Two years of additional development are required for this sensor to be compatible in performance with presently available Deft sensors.

Some aspects of current Deft sensor development could be utilized in improving the performance of the bulk acoustic wave (BAW) light valve. For example, a segmented transducer is currently under development which will produce wider transducer bandwidths. This new design could be used to increase the bandwidth of the BAW device also.

Other developments would have to be undertaken however, which are specifically required for the BAW sensor. In order

to achieve wider transducer bandwidths, small transducers with higher center frequencies are required. In the experimental BAW sensors, transducers were glued onto the quartz cell. The smaller transducers would be thinner and more brittle so that it would be necessary to sputter transducers directly onto the quartz cell.

Standing waves in the sensor have been mentioned as a current problem. These waves result from the reflection of the BAW from the cell boundaries. A number of techniques could be used to absorb the unwanted acoustic energy. For example, a larger cell could be used with only part of the cell serving as the active sensor volume. The remainder of the cell would then be used as an absorbing volume for the BAW. The edges of the cell could be joined with an absorbing material. The edges might also be sandblasted or made jagged to trap the acoustic energy. Another technique would be to drill small, random holes in the non-active volume of the cell to scatter the BAW. This could be done using a laser.

Present sensors exhibit a number of other problems which are discussed in Section II. Two to three years are required to produce a device suitable for feature extraction application. Since this work is developmental, the probability of success is perhaps 0.5

### C. Deft Sensor

The experimental results of Section IV indicate that presently available Deft sensors are capable of limited feature recognition. These sensors are being improved through a program of active support. Hence, the probability is good that improved sensors will be available which will have a realistic and significant feature recognition capability. This section will detail the current limitations and the steps being taken to correct these limitations.

The magnitude of the Fourier transform of the three reference features shown in Figure 32 as computed by a Deft

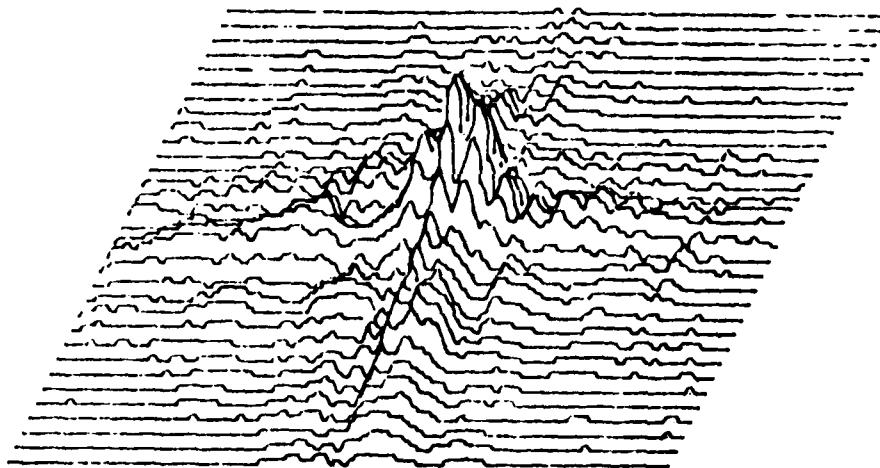
sensor are shown in Figures 34, 35 and 36. Figures 34 and 35 show the transform of two reference features in two different angular orientations. It can be seen that the transforms are (approximately) rotated versions of each other. Figure 36 shows the transform of a circle which was positioned at two locations on the sensor. These transforms are also (approximately) the same. The properties of translation invariance and transform rotation are required for all the algorithms developed in this report. Hence, the Deft sensor is potentially very useful for feature extraction. However, these properties are only approximately true for present sensors. Deviations in these properties as well as some other limitations restrict the usefulness of present Deft sensors in feature extraction. Limitations exist in the following categories:

1. spatial bandwidth
2. acoustic reflections
3. CdS uniformity
4. transform phase accuracy
5. output signal level

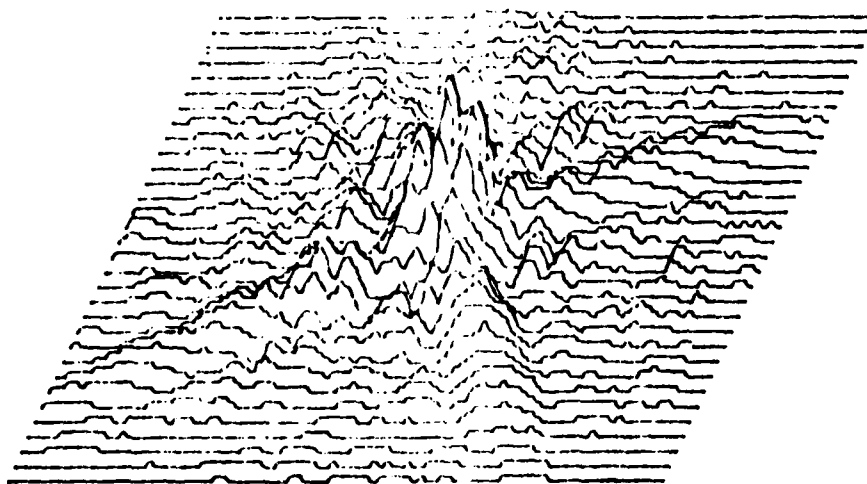
These categories will now be discussed one-by-one.

Present Deft sensors have spatial frequency bandwidths of  $\pm 10$  line pairs along one transform axis and  $\pm 20$  line pairs along the other axis. A larger bandwidth is desirable for the following three reasons:

1. To distinguish between features which are only subtly different from each other, high frequency information is required since the transforms may only differ in high frequency content.
2. Since the Fourier transform scales as the inverse of the feature scale, small features have large transforms. Large spatial bandwidths are needed to "see" large transforms. If bandwidth is limited then the image must be magnified. This requires breaking the photograph into a large number of views thereby increasing processing time.
3. If the scene contains "noise" which corrupts portions

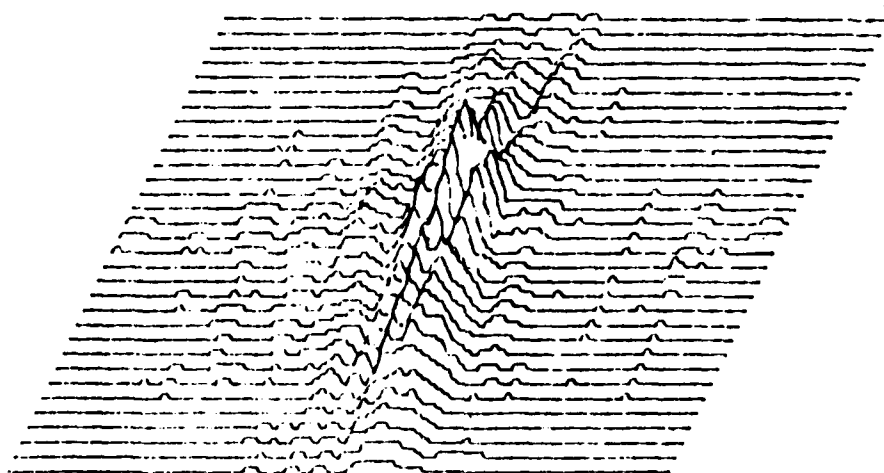


a.  $\theta = 0$  degrees

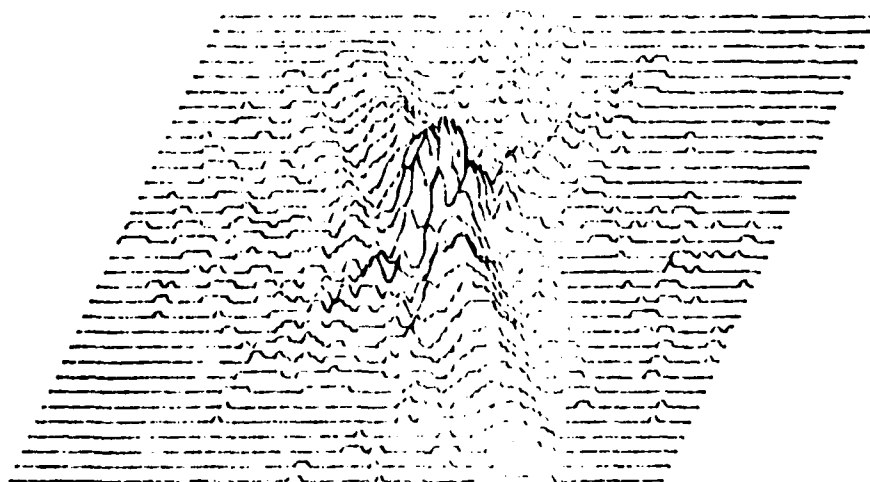


b.  $\theta = 25$  degrees

Figure 35 - Deft Transform of "Crossroads" Feature

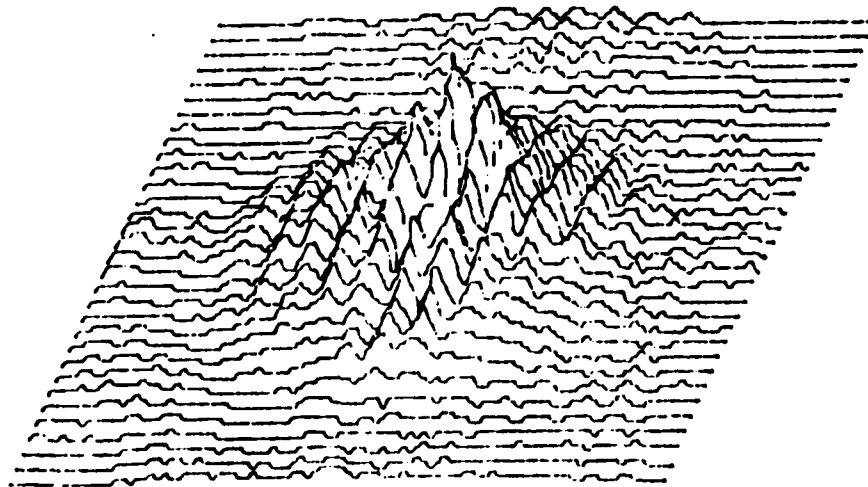


a.  $\theta = 0$  degrees

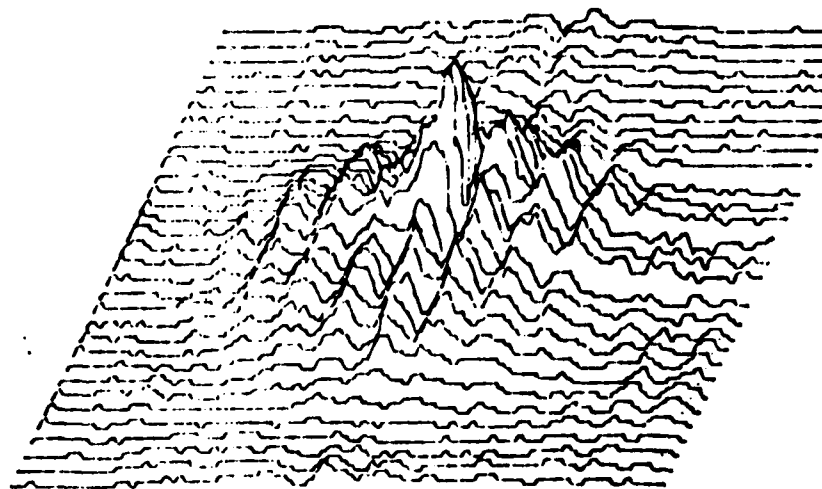


b.  $\theta = 25$  degrees

Figure 36 - Deft Transform of "Road" Feature



a. Circle at Center of Sensor



b. Circle at Edge of Sensor

Figure 37 - Deft Transform of Small Circle Feature



of the transform, a larger bandwidth may be required in order to have sufficient, useable transform components.

Deft sensors which will have larger bandwidths are under development. A sensor will be available by June 1981 which will have spatial frequency bandwidths of  $\pm 25$  line pairs along one transform axis and  $\pm 28$  line pairs along the other axis. This sensor will be able to "see" approximately four times the number of spatial frequency components as do current devices. This will be termed a "medium resolution" sensor. A "high resolution" sensor is also planned which should be available by January 1982. This sensor will have a fourfold increase in resolvable spatial frequencies over the medium resolution sensor.

In order to improve resolution, the medium resolution sensor will be fabricated on  $41.5^\circ$  rotated z-cut  $\text{LiNO}_3$ . The transducer center frequencies will also be increased from the present 35-40 MHz range to about 60 MHz. Increasing transducer center frequencies will result in a greater absolute bandwidth from a given percentage bandwidth. This is an attractive approach up to about 60MHz. Beyond this frequency, the mechanical loading of the metal grid pattern on the substrate leads to significant and undesirable damping of the SAW.

There are three advantages to using the new,  $41.5^\circ$  rotated z-cut  $\text{LiNO}_3$ . First, the coupling constants are larger than for the old cut. This means that more acoustic energy can be introduced into the sensor resulting in a larger output signal. Secondly, the spurious response of the sensor to a uniform image is smaller and further away from the main response peak than was the case for the old cut. A discussion of spurious acoustic modes in Deft sensors is found in Reference 31. If the sensor bandwidth were increased then the spurious would be within the bandwidth. Hence, it is important that the spurious is as small as possible. The third advantage is that with the rotated cut of  $\text{LiNO}_3$  the number of transducer finger pairs required for optimum energy coupling decreased from 8 to 6 in

one direction and from 8 to 4.5 in the other direction. Fewer finger pairs mean a larger percentage bandwidth.

The high resolution sensor will include these improvements. This sensor will also include an improved transducer design. The new transducers will be segmented in order to reduce their capacitance. Reduced capacitance allows wider transducer bandwidths. The resulting bandwidth leads to a significantly greater utility of Deft sensors in feature extraction.

One of the limitations of current Deft sensors is that these devices exhibit acoustic reflections from the regularly-spaced metal pick-up fingers. These reflections lead to a scalloping of the transform magnitude. This is evident in Figures 34, 35 and 36. The ripples which can be seen make feature extraction more difficult. For example, in Figure 37 by moving the circle from the center to the edge of the sensor the shape and the number of ripples changes. The feature has remained the same but the transform is really not invariant to translation which is a requirement for the preprocessor algorithms. Now refer to Figures 35 and 36. Notice that when the feature is in two different angular orientations the ripples may or may not appear on the "arms" of the transform. Hence, the transform is really not just rotated when the image rotates. There is also a shape change which is not desired. A third problem with these ripples occurs in the computation of image moments. The lower order moments are sensitive to the partial derivatives of the transform at the origin. Because of the ripples, these derivatives are drastically altered. In addition, they also become a function of feature position. In summary, ripples caused by reflections are very undesirable.

The medium resolution sensor will incorporate a new metal finger pattern spacing which will eliminate the reflections. This new spacing causes reflections from metal lines to add destructively. An important advantage of this spacing is that critical metal-to-metal dimensions are not reduced. In

general, device yield goes down as metal-to-metal dimensions are reduced because it is then more likely that contaminants will cause shorts across the metal lines.

A problem with current Deft sensors is that the CdS squares on the sensor do not conduct uniformly. Measurements have been made showing up to 2:1 variation in light conductivity over a sensor. This variation weights the image function. In the transform domain the image transform is convolved with the transform of the weight function. This tends to broaden spectral peaks and reduce or eliminate sidelobes. We typically observe peaks which are twice as wide as theory predicts. CdS variation leads to another problem. The computed transform may not be invariant to translation or rotate as the image rotates. This is evident in Figures 34, 35 and 36.

It has been determined that film variation is a result of the method used to deposit CdS on the substrate. The material, in the form of a gas, is passed over the substrate in a furnace. The substrate is almost as large as the furnace tube so that flow rates near the substrate edges are faster than at the middle. We presently have on order a new furnace tube with a larger inner diameter. Uniform flow rates should give uniform CdS film. An additional solution is to reduce the flow rate into the furnace in order to reduce turbulence. The success of these changes will be measured when the new equipment is in-place.

Another problem which was discussed in Section IV is the inaccuracy of the transform phase computed by the sensor. This is important in the method of moments. Part of the phase error must come from the same sources which produce errors in the transform magnitude. However, the sensor contains a phase term which is a function of the electronics (amplifiers, filters) and not the transform. These electronics were not originally designed to provide a linear phase response. Some recent measurements have been made which show that, indeed, the phase is nonlinear. Since these filters must be redesigned to

accommodate the new medium resolution sensor, they will be redesigned to exhibit approximate linear phase response.

The final item to be considered is output signal level. Present output signals are of the order of a microamp. In order to amplify this signal and achieve a good signal-to-noise ratio at the amplifier output, a rather narrow amplifier bandwidth is required. This narrow bandwidth affects the rate at which the sensor drive frequencies can be changed. If signal-to-noise ratio was not a problem then drive frequencies could be stepped every two microseconds which is the time required to propagate the SAW across the sensor. This would allow  $5 \times 10^5$  Fourier components to be addressed every second. But because of the need to narrow the output bandwidth, this rate is reduced to about  $10^3$  components per second. If the sensor output signal could be increased by a factor of 10 then  $10^4$  components per second would be possible.

The new medium resolution sensor will have an output signal approximately eight times as large as current sensors. This is because the  $41.5^\circ$  rotated z-cut  $\text{LiNO}_3$  results in better coupling of energy into the SAW.

All the factors which have been identified as limiting the performance of the Deft sensor in feature extraction applications have been listed and improvements under development discussed. In a number of cases, expected improvements can be stated with high probability of success. In other cases, improvements must be determined after new devices are fabricated and tested. The medium resolution sensor will be available by June 1981. The high resolution sensor will be available by January 1982. These sensors should allow significant improvements in feature extraction capability. It is suggested that these new sensors be examined for their applicability to feature extraction at the time they are available. This could be done with little cost by simply repeating the feature extraction experiments using the presently available computer programs and the new design sensors.

## VI. SUMMARY AND CONCLUSIONS

The objective of this program was to develop, analyze and evaluate theoretical concepts and strategies for topographic feature extraction and image analysis using acousto-optic (A-O) technology. To provide a frame of reference, a general feature extraction system model was developed in Section I. The most computation-intensive portion of the feature extraction process is the prefilter function. The purpose of this function is to reduce the large information content of the image to a much smaller set of values which can then be input to the decision processor. An important conclusion of this study was that A-O devices are potentially capable of implementing the prefilter function very efficiently. The input/output function of these devices is defined by equation (1). The device input is an image which is then transformed, using equation (1), into an electrical signal which can be input to a digital decision processor. The A-O devices have a number of desirable properties in this application. First of all, the device input/output format is ideal for the application. Secondly, lasers or precise optical alignments are not required. Third, these devices are rugged and potentially inexpensive. Finally, the function defined by equation (1) is central to a number of promising prefilter algorithms.

Transform-based prefilter algorithms were examined in Section III. Algorithms were developed there which are invariant to feature translation, rotation and scale. This invariance is highly desirable since it reduces the number of distinct feature signatures which must be processed by the decision processor. These algorithms require, as input, either the image two-dimensional Fourier transform or the image two-dimensional moments. Either of these functions can be computed efficiently using A-O devices.

Some preliminary experiments were conducted using the Fourier-based algorithms, test images and an A-O device which

was a Deft sensor. These experiments verified the invariance properties of the algorithms. The microcomputer-based Deft system was able to distinguish between three test patterns which were presented to the system in arbitrary orientation and scale. The success rate was 80%.

In spite of these promising results, present Deft sensors are not capable of distinguishing realistic features in aerial photographs. Present sensor limitations are identified and discussed in Section V. New Deft sensors are presently under development which will significantly improve the capability of this sensor in feature extraction applications. The expected improvements in a number of parameters can be predicted fairly accurately. However, other parameters must be measured after the new devices are available. It is suggested that the best way of determining the applicability of new sensors to feature extraction is to rerun the experiments reported here using the already written computer programs but using the new sensors. Since the microprocessor-based Deft system is already in-place and the computer programs written, these investigations could be accomplished with only a few man-weeks of effort.

## VII. REFERENCES

1. M. D. Levine, "Feature Extraction: A Survey," Proc. IEEE, Vol. 57, No. 8, pp. 1391-1407, Aug. 1969.
2. P. Kornreich, S. Kowel, A. Mahapatra and M. Mehter, "Monolithic Two-Dimensional Fourier Transformer," 1978 Int. Optical Computing Conference, pp. 49-54.
3. S. Kowel, P. Kornreich, A. Mahapatra, T. Szebenyi, A. Nouhi, "A Programmable, Multi-Function Processor," S.P.I.E. Tech. Symp. East, Washington, D.C., April 1979.
4. S. Kowel, "The Deft Camera," Optical Spectra, July 1980.
5. S. Kowel, D. Cleverly, A. Mahapatra, T. Szebenyi, P. Kornreich, "A Two-Dimensional Acoustic Processor," Int. Optical Computing Conference, pp. 35-39, 1978.
6. S. Kowel, D. Cleverly, A. Mahapatra, T. Szebenyi, P. Kornreich, S. Wanuga, "Two-Dimensional Photoacoustic Image Processing With Longitudinal Waves," Proc. Ultrasonics Symp. pp. 258-262, 1978.
7. H. Gautier, P. Ledu and C. Maerfeld, "Two-Dimensional Acoustic Fourier Transform of Optical Images," Proc. Ultrasonics Symp. 1979.
8. G. Nagy, "State of the Art in Pattern Recognition," Proc. IEEE, Vol. 56, No. 5, pp. 836-862, 1968.
9. S. Dudani, K. Breeding and R. McGhee, "Aircraft Identification by Moment Invariants," IEEE Trans. Computers, Vol. C-26, No. 1, pp. 39-45, 1977.
10. J. McLaughlin and J. Raviv, "Nth Order Autocorrelations in Pattern Recognition," IEEE Int. Symp. on Information Theory, Aug. 1967.
11. R. Pickholtz, "Investigation of Linear Transformations for Automatic Cartographic Analysis," Interim reports No. 1, 2 and 3, Contract No. DAAK 70-78-C-0045, 1978, 1979.
12. M. Hu, "Visual Pattern Recognition by Moment Invariants," IRE Trans. on Information Theory, pp. 179-187, Feb. 1962.
13. M. Teague, "Image Analysis via the General Theory of Moments," J. Opt. Soc. Am., Vol. 70, No. 8, pp. 920-930, Aug. 1980.
14. P. Chen and W. Seemuller, "Signal Signatures of Topographic Features Using Analog Technology," Report No. ETL-0185, U.S. Army Engineer Topographic Laboratories, Ft. Belvoir, Va. 1979.

15. G. Schreiber, "Bandwidth Requirements for Walsh Functions," Proc. Walsh Function Symp., pp. 46-51, 1970.
16. A. VanderLugt, "Signal Detection by Complex Spatial Filtering," IEEE Trans. on Information Theory, Vol. IT-10, pp. 139-145, April 1964.
17. A. VanderLugt, "Coherent Optical Processing," Proc. IEEE, Vol. 62, No. 10, pp. 1300-1319, Oct 1974.
18. S. Craig and A. Moyer, "Image Alignment and Correlation System," Final Report No. ETL-0237 for U.S. Army Contract DAAK70-78-C-0217, July 1980.
19. D. Casasent and A. Furman, "Sources of Correlation Degradation," Applied Optics, Vol. 16, No. 6, pp. 1652-1661, June 1977.
20. G. Lendaris and G. Stanley, "Diffraction-Pattern Sampling for Automatic Pattern Recognition," Proc. IEEE, Vol. 58, No. 2, pp. 198-216, Feb. 1970.
21. B. Pernick, R. Kopp, J. Lisa, J. Mendelsohn, H. Stone and P. Wohlers, "Screening of Cervical Cytological Samples Using Coherent Optical Processing. Part 1," Applied Optics, Vol. 17, No. 1, pp. 21-34, Jan. 1978.
22. R. Gordon and G. Herman, "Three-Dimensional Reconstruction from Projections: A Review of Algorithms," Int. Rev. of Cytology, Vol. 38, pp. 111-151, 1974.
23. D. Casasent and D. Psaltis, "New Optical Transforms for Pattern Recognition," Proc. IEEE, Vol. 65, No. 1, pp. 77-84, Jan. 1977.
24. D. Casasent and D. Psaltis, "Position, Rotation and Scale Invariant Optical Correlation," Applied Optics, Vol. 15, No. 7, pp. 1795-1799, July 1976.
25. D. Casasent and M. Kraus, "Polar Camera for Space-Variant Pattern Recognition," Applied Optics, Vol. 17, No. 10, pp. 1559-1561, May 1978.
26. D. Psaltis and D. Casasent, "Deformation Invariant Optical Processors Using Coordinate Transformations," Applied Optics, Vol. 16, No. 8, pp. 2288-2292, Aug. 1977.
27. D. Casasent and D. Psaltis, "Multiple-Invariant Space - Variant Optical Processors," Applied Optics, Vol. 17, No. 4, pp. 655-659, Feb. 1978.



28. J. Volder, "The Cordic Trigonometric Computing Technique," IRE Trans. Comp., pp. 330-334, Sept. 1959.
29. M. Teague, "Optical Calculation of Irradiance Moments," Applied Optics, Vol. 19, No. 8, pp. 1353-1356, April 1980.
30. E. Isaacson and H. Keller, Analysis of Numerical Methods, John Wiley and Sons, 1966, Chapter 5, Section 5.1.
31. A. Mahapatra, S. Kowel, P. Kornreich, M. Mehter, "Spurious Acoustic Modes in Two-Dimensional Fourier Transform Devices," IEEE Trans. on Sonics and Ultrasonics, Vol, SU-25, No. 6, Nov. 1978.

APPENDIX A - Method of Invariant Fourier Signatures  
Assembly Code Listing

This Appendix consists of a listing of the assembly language program which computes invariant Fourier signatures. This program was written to run on the Deft Laboratories' microprocessor-based test bed. All addresses and opcodes are hexadecimal. In the operand column of the statements the following symbols are used:

\$	Hexadecimal Prefix
%	Binary Prefix
H	Hexadecimal Postfix
D	Decimal Postfix
B	Binary Postfix
#	Denotes Immediate Addressing Mode

The entry address for this program is \$2000.

# METHOD OF INVARIANT FOURIER SIGNATURES

1800		ORG	\$1800
1800	TEMP1	RMB	1
1801	TEMP2	RMB	1
1802	TEMP3	RMB	1
1803	TEMP4	RMB	1
1804	TEMP5	RMB	1
1805	TEMP6	RMB	1
1806	BCD1	RMB	1
1807	BCD2	RMB	1
1824		ORG	\$1824
1824	COR1	RMB	1
1825	COR2	RMB	1
1826	COR3	RMB	1
1827	COR4	RMB	3
182A	COR9	RMB	1
182B	COR10	RMB	5
1830	UI1	RMB	1
1831	UI2	RMB	1
1832	VJ1	RMB	1
1833	VJ2	RMB	2
1835	DTIME	RMB	1
183D		ORG	\$183D.
183D	NSAMP	RMB	1
183E	LOGS	RMB	18
1850	PUSHST	RMB	1
18B3		ORG	\$18B3
18B3	FFTN	RMB	2
18E0		ORG	\$18E0
18E0	REF	RMB	8
18E8	FIRST	RMB	1
18E9	CIR	RMB	1
18EA	M	RMB	1
18EB	N	RMB	1
18EC	I	RMB	1
18ED	P	RMB	1
18EE	TAG	RMB	1
18EF	THETA	RMB	2
18F1	DTHETA	RMB	2
18F3	PHI	RMB	2
18F5	DPHI	RMB	2
18F7	STACKS	RMB	2
18F9	RUN	RMB	2
18FB	RUN1	RMB	2
18FD	STOP	RMB	2
18FF	INT	RMB	4
1903	S1	RMB	8
190B	S2	RMB	8
1913	S3	RMB	8
191B	AUG	RMB	6
1921	POS	RMB	3
1924	NEG	RMB	3
2A3F	PASC	EQU	\$2A3F
2A17	IDEX	EQU	\$2A17
1385	ROT15	EQU	\$1385
1376	VECT15	EQU	\$1376
29AE	BI*BCD	EQU	\$29AE
2A4C	RDDEFT	EQU	\$2A4C

## METHOD OF INVARIANT FOURIER SIGNATURES

29FE		TUNE	EQU	\$29FE
2999		DELAY1	EQU	\$2999
292F		TRAPIN	EQU	\$292F
2960		TRAPFX	EQU	\$2960
1000		FFT	EQU	\$1000
2A2A		WRITE	EQU	\$2A2A
8000		MATH	EQU	\$8000
290C		PUSH42	EQU	\$290C
28EF		PUSH41	EQU	\$28EF
28A3		PULL4	EQU	\$28A3
2700		RCORR1	EQU	\$2700
277F		RCORR2	EQU	\$277F
0FFF		FFTVT	EQU	\$FFF
0000		R0	EQU	\$0
0000			ORG	\$0
0000	39		FCB	\$39
2000			ORG	\$2000
		*		
		*	FEATURE EXTRACTION PROGRAM	
		*	DEFT/PROJECTION/FFT METHOD	
2000	CE 18 E0	FEAT#1	LDX	\$REF CLEAR REF(I)
2003	C6 08		LDAB	\$8
2005	6F 00	FE1	CLR	0,X
2007	08		INX	
2008	5A		DECB	
2009	26 FA		BNE	FE1
200B	7F 18 E8		CLR	FIRST
200E	BD FD A6		JSR	\$FDA6
2011	CE 25 7D		LDX	\$LINE1
2014	C6 13		LDAB	\$19D
2016	BD 2A 3F		JSR	PASC
2019	BD FD A6		JSR	\$FDA6
201C	CE 25 90		LDX	\$LINE2
201F	C6 13		LDAB	\$19D
2021	BD 2A 3F		JSR	PASC
2024	BD FD A6		JSR	\$FDA6
2027	BD FD A6		JSR	\$FDA6
202A	CE 25 A3		LDX	\$LINE3
202D	C6 15		LDAB	\$21D
202F	BD 2A 3F		JSR	PASC
2032	BD FD 36		JSR	\$FD36 INCH
2035	84 0F		ANDA	\$0F
2037	B7 18 3E		STAA	LOGS
203A	C6 01		LDAB	\$1
203C	4A	FE2	DECA	
203D	2D 03		BLT	FE3
203F	58		ASLB	
2040	20 FA		BRA	FE2
2042	F7 18 3D	FE3	STAB	NSAMP
2045	BD FD A6	FE4	JSR	\$FDA6
2048	CE 25 B8		LDX	\$LINE4
204B	C6 08		LDAB	\$8
204D	BD 2A 3F		JSR	PASC
2050	BD FD 36		JSR	\$FD36 INCH
2053	5F		CLRB	
2054	81 59		CMPA	\$59
2056	26 01		BNE	FE5

# METHOD OF INVARIANT FOURIER SIGNATURES

2058	5C				INCB		YES
2059	F7	18	E9	FE5	STAB	CIR	
205C	7D	18	E8		TST	FIRST	
205F	27	02			BEQ	FE6	
2061	20	46			BRA	FE11	
2063	7D	18	E9	FE6	TST	CIR	
2066	27	12			BEQ	FE7	
2068	BD	FD	A6		JSR	\$FDA6	
206B	CE	25	C0		LDX	\$LINE6	
206E	C6	15			LDAB	\$21D	
2070	BD	2A	3F		JSR	PASC	
2073	86	05			LDAA	\$5	
2075	B7	18	EA		STAA	M	
2078	20	19			BRA	FE8	
207A	BD	FD	A6	FE7	JSR	\$FDA6	
207D	CE	25	C0		LDX	\$LINE6	
2080	C6	14			LDAB	\$20D	
2082	BD	2A	3F		JSR	PASC	
2085	BD	FD	36		JSR	\$FD36	INCH
2088	81	5A			CMPA	\$5A	
208A	27	EE			BEQ	FE7	
208C	84	0F			ANDA	\$F	
208E	27	EA			BEQ	FE7	
2090	B7	18	EA		STAA	M	
2093	36			FE8	PSHA		
2094	86	80			LDAA	\$80	
2096	5F				CLRB		
2097	30				TSX		
2098	6D	00		FE9	TST	0,X	
209A	27	06			BEQ	FE10	
209C	6A	00			DEC	0,X	
209E	47				ASRA		
209F	56				RORB		
20A0	20	F6			BRA	FE9	
20A2	B7	18	F5	FE10	STAA	DPHI	
20A5	F7	18	F6		STAB	DPHI+1	
20A8	32				PULA		
20A9	BD	FD	A6	FE11	JSR	\$FDA6	
20AC	CE	25	D5		LDX	\$LINE7	
20AF	C6	13			LDAB	\$19D	
20B1	BD	2A	3F		JSR	PASC	
20B4	BD	FD	36		JSR	\$FD36	INCH
20B7	81	5A			CMPA	\$5A	
20B9	27	14			BEQ	FE12A	
20BB	84	0F			ANDA	\$F	
20BD	B7	18	EB		STAA	N	
20C0	36				PSHA		
20C1	86	80			LDAA	\$80	
20C3	5F				CLRB		
20C4	30				TSX		
20C5	6D	00		FE12	TST	0,X	
20C7	27	0E			BEQ	FE13	
20C9	6A	00			DEC	0,X	
20CB	47				ASRA		
20CC	56				RORB		
20CD	20	F6			BRA	FE12	
20CF	7D	18	E9	FE12A	TST	CIR	

# METHOD OF INVARIANT FOURIER SIGNATURES

20D2	2E D5	BGT	FE11	
20D4	36	PSHA		
20D5	4F	CLRA		
20D6	5F	CLRB		
20D7	B7 18 F1	STAA	DTHETA	
20DA	F7 18 F2	STAB	DTHETA+1	
20DD	32	PULA		
20DE	BD FD A6	JSR	\$FDA6	
20E1	CE 25 E8	LDX	\$LINE8	
20E4	C6 0E	LDAB	\$14D	
20E6	BD 2A 3F	JSR	PASC	
20E9	BD FD 36	JSR	\$FD36	INCH
20EC	84 0F	ANDA	\$\$F	
20EE	B7 18 EE	STAA	TAG	
20F1	7F 18 F3	CLR	PHI	PHI=0
20F4	7F 18 F4	CLR	PHI+1	
20F7	CE C0 00	LDX	\$\$C000	
20FA	FF 18 EF	STX	THETA	THETA=-PI/2
20FD	7F 18 EC	CLR	I	
2100	CE 00 00	LDX	\$0	
2103	FF 19 1B	STX	AVG	
2106	FF 19 1D	STX	AVG+2	
2109	FF 19 1F	STX	AVG+4	
210C	BF 18 F7	STS	STACKS	
210F	8E 0F FF	LDS	\$FFTVT	
2112	CE 00 00	LDX	\$0	LOOP: EACH CONTOUR
2115	FF 18 FF	STX	INT	
2118	FF 19 01	STX	INT+2	
211B	7F 18 ED	CLR	P	
211E	7F 18 E8	CLR	FIRST	
2121	F6 18 EC	LDAB	I	LOOP: EACH SAMPLE
2124	4F	CLRA		
2125	58	ASLB		
2126	49	ROLA		
2127	CE 26 1A	LDX	\$RVECT	
212A	BD 2A 17	JSR	IDEX	
212D	A6 00	LDAA	0,X	GET R(I) RADIUS
212F	B7 18 24	STAA	COR1	
2132	A6 01	LDAA	1,X	
2134	B7 18 25	STAA	COR2	
2137	B6 18 EF	LDAA	THETA	ANGLE
213A	F6 18 F0	LDAB	THETA+1	
213D	FB 18 F4	ADDB	PHI+1	
2140	B9 18 F3	ADCA	PHI	
2143	8B 40	ADDA	\$\$40	ADJUST QUADRANT
2145	2B 10	BMI	FE16	
2147	B7 18 00	STAA	TEMP1	
214A	F7 18 01	STAB	TEMP2	
214D	4F	CLRA		
214E	5F	CLRB		
214F	F0 18 01	SUBB	TEMP2	
2152	B2 18 00	SBCA	TEMP1	
2155	20 09	BRA	FE17	
2157	5D	TSTB		
2158	26 06	BNE	FE17	
215A	81 80	CHPA	\$\$80	
215C	26 02	BNE	FE17	

# METHOD OF INVARIANT FOURIER SIGNATURES

215E	4F			CLRA	
215F	5F			CLRB	
2160	8B	C0	FE17	ADDA	##C0
2162	B7	18	2A	STAA	COR9
2165	F7	18	2B	STAB	COR10
2168	7F	18	26	CLR	COR3
216B	7F	18	27	CLR	COR4
216E	BD	13	85	JSR	ROT15
2171	B6	18	24	LDAA	COR1
2174	F6	18	25	LDAB	COR2
2177	CE	00	05	LDX	#5
217A	47		FE18	ASRA	
217B	56			RORB	
217C	09			DEX	
217D	26	FB		BNE	FE18
217F	FB	26	17	ADDB	XZERO+1
2182	B9	26	16	ADCA	XZERO
2185	B7	18	02	STAA	TEMP3
2188	F7	18	03	STAB	TEMP4
218B	BD	29	AE	JSR	BI\$BCD
218E	FE	18	06	LDX	BCD1
2191	FF	18	30	STX	UI1
2194	B6	18	26	LDAA	COR3
2197	F6	18	27	LDAB	COR4
219A	CE	00	05	LDX	#5
219D	47		FE19	ASRA	
219E	56			RORB	
219F	09			DEX	
21A0	26	FB		BNE	FE19
21A2	FB	26	19	ADDB	YZERO+1
21A5	B9	26	18	ADCA	YZERO
21A8	B7	18	02	STAA	TEMP3
21AB	F7	18	03	STAB	TEMP4
21AE	BD	29	AE	JSR	BI\$BCD
21B1	FE	18	06	LDX	BCD1
21B4	FF	18	32	STX	VJ1
21B7	7D	18	E8	TST	FIRST
21BA	27	0E		BEQ	FE19A
21BC	BD	2A	4C	JSR	RDDEFT
21BF	BD	13	76	JSR	VECT15
21C2	BD	29	2F	JSR	TRAPIN
21C5	20	06		BRA	FE19B
21C7	7E	21	21	JMP	FE15
21CA	7C	18	E8	INC	FIRST
21CD	BD	29	FE	JSR	TUNE
21D0	86	00		LDAA	#0
21D2	B7	18	35	STAA	DTIME
21D5	BD	29	99	JSR	DELAY1
21D8	7D	18	E9	TST	CIR
21DB	27	44		BEQ	FE20
21DD	B6	18	EF	LDAA	THETA
21E0	F6	18	F0	LDAB	THETA+1
21E3	FB	18	F2	ADDB	DTHETA+1
21E6	B9	18	F1	ADCA	DTHETA
21E9	B7	18	EF	STAA	THETA
21EC	F7	18	F0	STAB	THETA+1
21EF	7D	18	F0	TST	THETA+1

POLAR TO RECTANGULAR

MAG  
INTEGRATE

CIR=1

THETA=THETA+DTHETA

# METHOD OF INVARIANT FOURIER SIGNATURES

21F2	26	D3		BNE	FE19C	
21F4	B6	18	EF	LDAA	THETA	
21F7	8B	C0		ADDA	##C0	
21F9	26	CC		BNE	FE19C	
21FB	BD	29	60	JSR	TRAPFX	THETA=+PI/2
21FE	B6	19	02	LDAA	INT+3	
2201	36			PSHA		
2202	B6	19	01	LDAA	INT+2	
2205	36			PSHA		
2206	B6	19	00	LDAA	INT+1	
2209	36			PSHA		
220A	B6	18	FF	LDAA	INT	
220D	36			PSHA		
220E	B6	18	EC	LDAA	I	
2211	81	1F		CMPA	##31D	
2213	27	6A		BEG	FE23	
2215	7C	18	EC	INC	I	
2218	CE	C0	00	LDX	##C000	
221B	FF	18	EF	STX	THETA	
221E	7E	21	12	JMP	FE14	
2221	B6	18	EC	LDAA	I	FE20
2224	81	1F		CMPA	##31D	
2226	27	18		BEG	FE21	
2228	7C	18	EC	INC	I	
222B	B6	18	EF	LDAA	THETA	
222E	F6	18	F0	LDAB	THETA+1	
2231	FB	18	F2	ADDB	DTHETA+1	
2234	B9	18	F1	ADCA	DTHETA	
2237	B7	18	EF	STAA	THETA	
223A	F7	18	F0	STAB	THETA+1	
223D	7E	21	21	JMP	FE15	
2240	BD	29	60	JSR	TRAPFX	I=31
2243	B6	19	02	LDAA	INT+3	
2246	36			PSHA		
2247	B6	19	01	LDAA	INT+2	
224A	36			PSHA		
224B	B6	19	00	LDAA	INT+1	
224E	36			PSHA		
224F	B6	18	FF	LDAA	INT	
2252	36			PSHA		
2253	B6	18	F3	LDAA	PHI	
2254	F6	18	F4	LDAB	PHI+1	
2259	FB	18	F6	ADDB	DPHI+1	
225C	B9	18	F5	ADCA	DPHI	
225F	B7	18	F3	STAA	PHI	PHI=PHI+DPHI
2262	F7	18	F4	STAB	PHI+1	
2265	7D	18	F4	TST	PHI+1	
2268	26	09		BNE	FE22	
226A	B6	18	F3	LDAA	PHI	
226D	80	80		SUBA	##80	
226F	26	02		BNE	FE22	
2271	20	10		BRA	FE24	
2273	CE	C0	00	LDX	##C000	FE22
2276	FF	18	EF	STX	THETA	
2279	7F	18	EC	CLR	I	
227C	7E	21	12	JMP	FE14	



# METHOD OF INVARIANT FOURIER SIGNATURES

227F	B6 05	FE23	LDAA	#5	
2281	20 03		BRA	FE25	
2283	B6 18 EA	FE24	LDAA	M	
2286	B7 18 50	FE25	STAA	PUSHST	
2289	F6 19 1B	FE25A	LDAB	AVG	AVG=AVG/2**M
228C	57		ASRB		
228D	F7 19 1B		STAB	AVG	
2290	F6 19 1C		LDAB	AVG+1	
2293	56		RORB		
2294	F7 19 1C		STAB	AVG+1	
2297	F6 19 1D		LDAB	AVG+2	
229A	56		RORB		
229B	F7 19 1D		STAB	AVG+2	
229E	F6 19 1E		LDAB	AVG+3	
22A1	56		RORB		
22A2	F7 19 1E		STAB	AVG+3	
22A5	F6 19 1F		LDAB	AVG+4	
22A8	56		RORB		
22A9	F7 19 1F		STAB	AVG+4	
22AC	F6 19 20		LDAB	AVG+5	
22AF	56		RORB		
22B0	F7 19 20		STAB	AVG+5	
22B3	4A		DECA		
22B4	26 D3		BNE	FE25A	
22B6	30		TSX		
22B7	7F 19 21		CLR	POS	
22BA	7F 19 22		CLR	POS+1	
22BD	7F 19 23		CLR	POS+2	
22C0	B6 FF		LDAA	##FF	
22C2	B7 19 24		STAA	NEG	
22C5	B7 19 25		STAA	NEG+1	
22C8	B7 19 26		STAA	NEG+2	
22CB	A6 03	FE25B	LDAA	3,X	REMOVE MEAN
22CD	B0 19 20		SUBA	AVG+5	
22D0	A7 03		STAA	3,X	
22D2	A6 02		LDAA	2,X	
22D4	B2 19 1F		SBCA	AVG+4	
22D7	A7 02		STAA	2,X	
22D9	A6 01		LDAA	1,X	
22DB	B2 19 1E		SBCA	AVG+3	
22DE	A7 01		STAA	1,X	
22E0	A6 00		LDAA	0,X	
22E2	B2 19 1D		SBCA	AVG+2	
22E5	A7 00		STAA	0,X	
22E7	2B 18		BMI	FE25C	FIND SCALE FACTOR
22E9	BA 19 21		ORAA	POS	
22EC	B7 19 21		STAA	POS	
22EF	A6 01		LDAA	1,X	
22F1	BA 19 22		ORAA	POS+1	
22F4	B7 19 22		STAA	POS+1	
22F7	A6 02		LDAA	2,X	
22F9	BA 19 23		ORAA	POS+2	
22FC	B7 19 23		STAA	POS+2	
22FF	20 16		BRA	FE25D	
2301	B4 19 24	FE25C	ANDA	NEG	
2304	B7 19 24		STAA	NEG	
2307	A6 01		LDAA	1,X	

# METHOD OF INVARIANT FOURIER SIGNATURES

2309	B4 19 25		ANDA	NEG+1	
230C	B7 19 25		STAA	NEG+1	
230F	A6 02		LDAA	2,X	
2311	B4 19 26		ANDA	NEG+2	
2314	B7 19 26		STAA	NEG+2	
2317	08	FE25D	INX		
2318	08		INX		
2319	08		INX		
231A	08		INX		
231B	8C 10 00		CPX	#FFTUT+1	
231E	26 AB		BNE	FE25B	
2320	B6 19 24		LDAA	NEG	
2323	F6 19 25		LDAB	NEG+1	
2326	78 19 26		ASL	NEG+2	
2329	CE 00 00		LDX	#0	
232C	59	FE25E	ROLB		
232D	49		ROLA		
232E	24 04		BCC	FE25F	
2330	08		INX		
2331	0C		CLC		
2332	20 F8		BRA	FE25E	
2334	FF 19 24	FE25F	STX	NEG	
2337	B6 19 21		LDAA	POS	
233A	F6 19 22		LDAB	POS+1	
233D	78 19 23		ASL	POS+2	
2340	CE 00 00		LDX	#0	
2343	59	FE25G	ROLB		
2344	49		ROLA		
2345	25 04		BCS	FE25H	
2347	08		INX		
2348	0D		SEC		
2349	20 F8		BRA	FE25G	
234B	FF 19 21	FE25H	STX	POS	
234E	B6 19 22		LDAA	POS+1	
2351	B0 19 25		SUBA	NEG+1	
2354	2B 06		BMI	FE25I	
2356	B6 19 25		LDAA	NEG+1	
2359	B7 19 22		STAA	POS+1	
235C	86 11	FE25I	LDAA	#17D	
235E	B0 19 22		SUBA	POS+1	
2361	2F 27		BLE	FE25L	
2363	B7 19 22		STAA	POS+1	
2366	30		TSX		
2367	F6 19 22	FE25J	LDAB	POS+1	SCALE DOWN
236A	A6 00	FE25K	LDAA	0,X	
236C	47		ASRA		
236D	A7 00		STAA	0,X	
236F	A6 01		LDAA	1,X	
2371	46		RORA		
2372	A7 01		STAA	1,X	
2374	A6 02		LDAA	2,X	
2376	46		RORA		
2377	A7 02		STAA	2,X	
2379	A6 03		LDAA	3,X	
237B	46		RORA		
237C	A7 03		STAA	3,X	
237E	5A		DECB		

# METHOD OF INVARIANT FOURIER SIGNATURES

237F	26 E9		BNE	FE25K	
2381	08		INX		
2382	08		INX		
2383	08		INX		
2384	08		INX		
2385	8C 10 00		CPX	#FFTUT+1	
2388	26 DD		BNE	FE25J	
238A	B6 18 50	FE25L	LDAA	PUSHST	
238D	5F		CLRB		
238E	BD 10 00		JSR	FFT	FFT
2391	B6 18 B3		LDAA	FTTN	
2394	F6 18 B4		LDAB	FTTN+1	
2397	47		ASRA		
2398	56		RORB		
2399	CE 00 00		LDX	#R0	
239C	BD 2A 17		JSR	IDEX	
239F	FF 18 FD		STX	STOP	
23A2	CE 00 00		LDX	#R0	
23A5	FF 18 F9		STX	RUN	
23A8	CE 18 24	FE26	LDX	#COR1	PULL FFT/MAG/STORE
23AB	32		PULA		
23AC	A7 00		STAA	0,X	
23AE	32		PULA		
23AF	A7 01		STAA	1,X	
23B1	32		PULA		
23B2	A7 02		STAA	2,X	
23B4	32		PULA		
23B5	A7 03		STAA	3,X	
23B7	BD 13 76		JSR	VECT15	
23BA	FE 18 F9		LDX	RUN	
23BD	B6 18 24		LDAA	COR1	
23C0	A7 00		STAA	0,X	
23C2	B6 18 25		LDAA	COR2	
23C5	A7 01		STAA	1,X	
23C7	08		INX		
23C8	08		INX		
23C9	FF 18 F9		STX	RUN	
23CC	BC 18 FD		CPX	STOP	
23CF	26 D7		BNE	FE26	
23D1	BE 18 F7		LDS	STACKS	
23D4	BD FD A6		JSR	\$FDA6	
23D7	CE 25 F6		LDX	\$LINE9	
23DA	C6 0B		LDAB	\$IID	
23DC	BD 2A 3F		JSR	PASC	
23DF	BD FD 36		JSR	\$FD36	INCH
23E2	81 59		CHPA	\$59	
23E4	26 50		BNE	FE29	
23E6	CE 00 00		LDX	#R0	
23E9	FF 18 F9		STX	RUN	
23EC	86 01		LDAA	#1	
23EE	B7 18 EC		STAA	I	
23F1	BD FD A6		JSR	\$FDA6	
23F4	BD FD A6		JSR	\$FDA6	
23F7	FE 18 F9	FE27	LDX	RUN	PRINT SIGNATURE
23FA	A6 00		LDAA	0,X	
23FC	B7 18 02		STAA	TEMP3	
23FF	A6 01		LDAA	1,X	

# METHOD OF INVARIANT FOURIER SIGNATURES

2401	B7 18 03	STAA	TEMP4	
2404	08	INX		
2405	08	INX		
2406	FF 18 F9	STX	RUN	
2409	BD 29 AE	JSR	BI\$BCD	
240C	CE 18 06	LDX	#BCD1	
240F	BD 2A 2A	JSR	WRITE	
2412	CE 18 07	LDX	#BCD2	
2415	BD 2A 2A	JSR	WRITE	
2418	86 20	LDAA	#\$20	
241A	BD FD 80	JSR	\$FD80	BLANK
241D	BD FD 80	JSR	\$FD80	
2420	86 18 EC	LDAA	I	
2423	48	ABLA		
2424	24 05	BCC	FE28	
2426	BD FD A6	JSR	\$FDA6	
2429	86 01	LDAA	#1	
242B	B7 18 EC FE28	STAA	I	
242E	FE 18 F9	LDX	RUN	
2431	BC 18 FD	CPX	STOP	
2434	26 C1	BNE	FE27	
2436	BD FD A6 FE29	JSR	\$FDA6	
2439	BD FD A6	JSR	\$FDA6	
243C	CE 26 01	LDX	#LINE10	
243F	C6 0B	LDAB	#11D	
2441	BD 2A 3F	JSR	PASC	
2444	BD FD A6	JSR	\$FDA6	
2447	86 01	LDAA	#1	
2449	B7 18 EC	STAA	I	
244C	B6 18 EC FE30	LDAA	I	
244F	B1 18 EE	CHPA	TAG	
2452	26 53	BNE	FE33	
2454	16	TAB		
2455	8D 02	BSR	FE30A	
2457	20 20	BRA	FE32	
2459	B6 18 EA FE30A	LDAA	M	
245C	B7 18 00	STAA	TEMP1	
245F	4F	CLRA		
2460	58 FE31	ASLB		
2461	49	ROLA		
2462	7A 18 00	DEC	TEMP1	
2465	26 F9	BNE	FE31	
2467	58	ASLB		
2468	49	ROLA		
2469	CE 00 00	LDX	#R0	
246C	BD 2A 17	JSR	IDEX	X=R0+I*2***(M+1)
246F	FF 18 FB	STX	RUN1	
2472	CE 00 00	LDX	#R0	
2475	FF 18 F9	STX	RUN	
2478	39	RTS		
2479	FE 18 F9 FE32	LDX	RUN	STORE NEW SIGNATURE IN R(1)
247C	A6 00	LDAA	0,X	
247E	E6 01	LDAB	1,X	
2480	08	INX		
2481	08	INX		
2482	FF 18 F9	STX	RUN	
2485	FE 18 FB	LDX	RUN1	

# METHOD OF INVARIANT FOURIER SIGNATURES

2488	A7 00	STAA	0,X	
248A	E7 01	STAB	1,X	
248C	08	INX		
248D	08	INX		
248E	FF 18 FB	STX	RUN1	
2491	FE 18 F9	LDX	RUN	
2494	BC 18 FD	CPX	STOP	
2497	26 E0	BNE	FE32	
2499	CE 18 DF	LDX	#REF-1	
249C	F6 18 EC	LDAB	I	
249F	4F	CLRA		
24A0	BD 2A 17	JSR	IDEX	
24A3	86 01	LDAA	#1	
24A5	A7 00	STAA	0,X	SET REF(I)=1
24A7	F6 18 EC FE33	LDAB	I	
24AA	4F	CLRA		
24AB	CE 18 DF	LDX	#REF-1	
24AE	BD 2A 17	JSR	IDEX	
24B1	A6 00	LDAA	0,X	
24B3	81 01	CMPA	#1	
24B5	27 09	BEQ	FE34	
24B7	7F 18 06	CLR	BCD1	
24BA	7F 18 07	CLR	BCD2	
24BD	7E 25 14	JMP	FE37	
24C0	F6 18 EC FE34	LDAB	I	REF(I)=1
24C3	8D 94	BSR	FE30A	
24C5	4F	CLRA		
24C6	C6 18	LDAB	#24D	
24C8	CE 19 03	LDX	#S1	
24CB	A7 00 FE35	STAA	0,X	S1,S2,S3 NFPZ
24CD	08	INX		
24CE	5A	DECB		
24CF	26 FA	BNE	FE35	
24D1	86 80	LDAA	#80	
24D3	B7 19 0A	STAA	S1+7	
24D6	B7 19 12	STAA	S2+7	
24D9	B7 19 1A	STAA	S3+7	
24DC	FE 18 F9 FE36	LDX	RUN	CORRELATE R(0) WITH R(I)
24DF	A6 00	LDAA	0,X	
24E1	B7 18 02	STAA	TEMP3	
24E4	A6 01	LDAA	1,X	
24E6	B7 18 03	STAA	TEMP4	
24E9	08	INX		
24EA	08	INX		
24EB	FF 18 F9	STX	RUN	
24EE	FE 18 FB	LDX	RUN1	
24F1	A6 00	LDAA	0,X	
24F3	B7 18 04	STAA	TEMP5	
24F6	A6 01	LDAA	1,X	
24F8	B7 18 05	STAA	TEMP6	
24FB	08	INX		
24FC	08	INX		
24FD	FF 18 FB	STX	RUN1	
2500	CE 18 02	LDX	#TEMP3	
2503	BD 27 00	JSR	RCORR1	ACCUMULATE
2506	FE 18 F9	LDX	RUN	
2509	BC 18 FD	CPX	STOP	

# METHOD OF INVARIANT FOURIER SIGNATURES

250C	26	CE		BNE	FE36	
250E	BD	27	7F	JSR	RCORR2	81/(82*93)**0.5
2511	BD	29	AE	JSR	BI*BCD	
2514	B6	18	EC	LDAA	I	
2517	8A	30		ORAA	**30	
2519	BD	FD	80	JSR	\$FD80	PRINT I
251C	86	29		LDAA	**29	
251E	BD	FD	80	JSR	\$FD80	PRINT ')
2521	86	20		LDAA	**20	
2523	BD	FD	80	JSR	\$FD80	BLANK
2526	86	30		LDAA	**30	
2528	BD	FD	80	JSR	\$FD80	PRINT 0
252B	86	2E		LDAA	**2E	
252D	BD	FD	80	JSR	\$FD80	PRINT '.,'
2530	CE	18	06	LDX	*BCD1	
2533	BD	2A	2A	JSR	WRITE	CORR
2536	CE	18	07	LDX	*BCD2	
2539	BD	2A	2A	JSR	WRITE	
253C	86	20		LDAA	**20	
253E	BD	FD	80	JSR	\$FD80	BLANK
2541	BD	FD	80	JSR	\$FD80	
2544	B6	18	EC	LDAA	I	
2547	81	04		CMPA	*4	
2549	26	03		BNE	FE38	
254B	BD	FD	A6	JSR	\$FDA6	
254E	7C	18	EC	INC	I	
2551	B6	18	EC	LDAA	I	
2554	81	09		CMPA	*9	
2556	26	1F		BNE	FE39	
2558	BD	FD	A6	JSR	\$FDA6	
255B	CE	26	0C	LDX	*LINE11	
255E	C6	0A		LDAB	*10D	
2560	BD	2A	3F	JSR	PASC	
2563	BD	FD	36	JSR	\$FD36	INCH
2566	81	59		CMPA	*59	
2568	26	10		BNE	FE40	
256A	B6	18	EA	LDAA	H	
256D	81	05		CMPA	*5	
256F	27	03		BEQ	FE38A	
2571	7E	20	A9	JMP	FE11	
2574	7E	20	45	JMP	FE4	
2577	7E	24	4C	JMP	FE30	
257A	7E	FE	2D	JMP	\$FE2D	FANTOM
257D			LINE1	EQU	*	
257D	46	45	41	FCC	'FEATURE '	
2585	52	45	43	FCC	'RECOGNITION'	
2590			LINE2	EQU	*	
2590	44	45	46	FCC	'DEFT/PROJECTION/'	
25A0	46	46	54	FCC	'FFT'	
25A3			LINE3	EQU	*	
25A3	53	41	4D	FCC	'SAMPLE AVERAGE'	
25B1	20	49	53	FCC	' IS 2**'	
25B8			LINE4	EQU	*	
25B8	43	49	52	FCC	'CIRCLES?'	
25C0			LINE6	EQU	*	
25C0	52	45	46	FCC	'REFERENCE LENGTH'	
25D0	3D	32	2A	FCC	'=2**5'	

# METHOD OF INVARIANT FOURIER SIGNATURES

25D5		LINE7	EQU	*	
25D5	44 45 4C		FCC	'DELTA THETA=-PI/'	
25E5	32 2A 2A		FCC	'2**'	
25E8		LINE8	EQU	*	
25E8	52 45 46		FCC	'REFERENCE TAG='	
25F6		LINE9	EQU	*	
25F6	53 49 47		FCC	'SIGNATURE?'	
2601		LINE10	EQU	*	
2601	43 4F 52		FCC	'CORRELATION'	
260C		LINE11	EQU	*	
260C	43 4F 4E		FCC	'CONTINUE?'	
2616		XZERO	EQU	*	
2616	0F 11		FCB	\$F,\$11	3857
2618		YZERO	EQU	*	
2618	0D AA		FCB	\$D,\$AA	3498
261A		RVECT	EQU	*	
261A	07 2B 07		FCB	\$7,\$2B,\$7,\$C8,\$8,\$72,\$9,\$2B	
2622	09 F4 0A		FCB	\$9,\$F4,\$A,\$CE,\$B,\$BA,\$C,\$BB	
262A	0D D1 0E		FCB	\$D,\$D1,\$E,\$FF,\$10,\$47,\$11,\$AB	
2632	13 2E 14		FCB	\$13,\$2E,\$14,\$D1,\$16,\$99,\$18,\$87	
263A	1A A0 1C		FCB	\$1A,\$A0,\$1C,\$E6,\$1F,\$5F,\$22,\$D	
2642	24 F6 28		FCB	\$24,\$F6,\$28,\$1E,\$2B,\$8C,\$2F,\$45	
264A	33 4F 37		FCB	\$33,\$4F,\$37,\$B1,\$3C,\$74,\$41,\$9E	
2652	47 3A 4D		FCB	\$47,\$3A,\$4D,\$50,\$53,\$EB,\$5B,\$17	
265A			END		

STATEMENTS =710

FREE BYTES =16551

NO ERRORS DETECTED

# METHOD OF INVARIANT FOURIER SIGNATURES

18B0		ORG	\$18B0
18B0	BLKSF	RMB	1
18B1	M	RMB	1
18B2	TFLAG	RMB	1
18B3	N	RMB	2
18B5	LE	RMB	2
18B7	ANGLE	RMB	2
18B9	L	RMB	1
18BA	LE1	RMB	2
18BC	THETA	RMB	2
18BE	J	RMB	2
18C0	I	RMB	2
18C2	BUT1R	RMB	2
18C4	BUT1I	RMB	2
18C6	BUT2R	RMB	2
18C8	BUT2I	RMB	2
18CA	IP	RMB	2
18CC	II	RMB	2
18CE	NV2	RMB	2
18D0	NM1	RMB	2
18D2	K	RMB	2
18D4	JJ	RMB	2
18D6	BSN	RMB	2
18D8	POS	RMB	2
18DA	NEG	RMB	2
18DC	BSN1	RMB	2
1824		ORG	\$1824
1824	COR1	RMB	1
1825	COR2	RMB	1
1826	COR3	RMB	1
1827	COR4	RMB	1
1828	DNUP1	RMB	1
1829	DNUP2	RMB	1
182A	COR9	RMB	1
182B	COR10	RMB	1
1800		ORG	\$1800
1800	COR5	RMB	1
1801	COR6	RMB	1
1802	COR7	RMB	1
1803	COR8	RMB	1
1804	COR11	RMB	1
1805	COR12	RMB	1
1806	COR13	RMB	1
1807	COR14	RMB	1
1808	COR15	RMB	1
1809	COR16	RMB	1
180A	COR17	RMB	1
180B	COR18	RMB	1
180C	CFLAG	RMB	1
180D	COR19	RMB	2
180F	ITER	RMB	1
1000		ORG	\$1000

\* SUBROUTINE FFT  
 \* REAL, IMAG ALTERNATE IN STACK  
 \* ENTER WITH A ACCUM=M FOR  
 \* 2\*\*M FFT; B=0 FOR FORWARD FFT,  
 \* B=1 FOR INVERSE FFT



# METHOD OF INVARIANT FOURIER SIGNATURES

1000	7F 18 B0	FFT	CLR	BLKSF	
1003	CE 18 B1		LDX	#M	
1006	A7 00		STAA	0,X	M
1008	E7 01		STAB	1,X	TFLAG
100A	C6 01		LDAB	#1	
100C	4C		INCA		
100D	4C		INCA		
100E	0C		CLC		
100F	6F 02		CLR	2,X	
1011	59	F1	ROLB		
1012	24 04		BCC	F2	
1014	6F 03		CLR	3,X	
1016	09		DEX		
1017	5C		INCB		
1018	4A	F2	DECA		
1019	26 F6		BNE	F1	
101B	E7 03		STAB	3,X	N=2*(M+2)
101D	CE 18 B1		LDX	#M	
1020	A6 02		LDAA	2,X	
1022	E6 03		LDAB	3,X	
1024	58		ASLB		
1025	49		ROLA		
1026	A7 04		STAA	4,X	LE=2*(M+3)
1028	E7 05		STAB	5,X	
102A	A6 00		LDAA	0,X	
102C	C6 80		LDAB	#80	
102E	6F 07		CLR	7,X	
1030	0C		CLC		
1031	56	F3	RORB		
1032	24 04		BCC	F4	
1034	6F 06		CLR	6,X	
1036	08		INX		
1037	56		RORB		
1038	4A	F4	DECA		
1039	26 F6		BNE	F3	
103B	E7 06		STAB	6,X	ANGLE=2*(15-M)
103D	CE 18 B1		LDX	#M	
1040	6D 01		TST	1,X	
1042	2E 0A		BGT	F4A	
1044	4F		CLRA		FORWARD FFT
1045	5F		CLRB		
1046	E0 07		SUBB	7,X	
1048	A2 06		SBCA	6,X	
104A	A7 06		STAA	6,X	ANGLE=-ANGLE
104C	E7 07		STAB	7,X	
104E	86 01	F4A	LDAA	#1	
1050	A7 08		STAA	8,X	
1052	FE 18 B3	F5	LDX	N	MAIN LOOP
1055	B0 12 AC		JSR	BSCALE	
1058	BB 18 B0		ADDA	BLKSF	
105B	B7 18 B0		STAA	BLKSF	BLOCK FLOAT POINT SCALE F4
105E	CE 18 B5		LDX	#LE	
1061	A6 00		LDAA	0,X	
1063	E6 01		LDAB	1,X	
1065	44		LSRA		
1066	56		RORB		
1067	A7 00		STAA	0,X	LE=LE/2

# METHOD OF INVARIANT FOURIER SIGNATURES

1049	E7 01		STAB	1,X	
104B	44		LSRA		
104C	56		RORB		
104D	A7 05		STAA	5,X	
104F	E7 06		STAB	6,X	LE1=LE/2
1071	A6 02		LDAA	2,X	
1073	E4 03		LDAB	3,X	
1075	58		ASLB		
1076	49		ROLA		
1077	A7 02		STAA	2,X	ANGLE=2*ANGLE
1079	E7 03		STAB	3,X	
107B	6F 07		CLR	7,X	THETA=0
107D	6F 08		CLR	8,X	
107F	6F 09		CLR	9,X	
1081	86 04		LDAA	#4	
1083	A7 0A		STAA	10D,X	J=4
1085	30	F6	TSX		
1086	09		DEX		
1087	09		DEX		
1088	FF 18 C2		STX	BUT1R	
108B	CE 18 BE		LDX	#J	
108E	A6 00		LDAA	0,X	
1090	E6 01		LDAB	1,X	
1092	A7 0E		STAA	14D,X	II=J
1094	E7 0F		STAB	15D,X	
1096	EB 05		ADDB	5,X	
1098	A9 04		ADCA	4,X	
109A	A7 02		STAA	2,X	I=J+STKPTR-1
109C	E7 03		STAB	3,X	
109E	CE 18 BA F7		LDX	#LE1	
10A1	A6 00		LDAA	0,X	
10A3	E6 01		LDAB	1,X	
10A5	EB 07		ADDB	7,X	
10A7	A9 06		ADCA	6,X	
10A9	A7 10		STAA	16D,X	IP=I+LE1
10AB	E7 11		STAB	17D,X	
10AD	FE 18 C0		LDX	I	BUTTERFLY
10B0	A6 00		LDAA	0,X	
10B2	B7 18 C2		STAA	BUT1R	
10B5	A6 01		LDAA	1,X	
10B7	B7 18 C3		STAA	BUT1R+1	
10BA	A6 02		LDAA	2,X	
10BC	B7 18 C4		STAA	BUT1I	
10BF	A6 03		LDAA	3,X	
10C1	B7 18 C5		STAA	BUT1I+1	
10C4	FE 18 CA		LDX	IP	
10C7	A6 00		LDAA	0,X	
10C9	B7 18 C6		STAA	BUT2R	
10CC	A6 01		LDAA	1,X	
10CE	B7 18 C7		STAA	BUT2R+1	
10D1	A6 02		LDAA	2,X	
10D3	B7 18 C8		STAA	BUT2I	
10D6	A6 03		LDAA	3,X	
10D8	B7 18 C9		STAA	BUT2I+1	
10DB	CE 18 C2		LDX	#BUT1R	
10DE	A6 00		LDAA	0,X	
10E0	E6 01		LDAB	1,X	

ME METHOD OF INVARIANT FOURIER SIGNATURES

10E2	EB 05	ADDB	5,X
10E4	A9 04	ADCA	4,X
10E6	B7 18 24	STAA	COR1
10E9	F7 18 25	STAB	COR2
10EC	A6 02	LDAA	2,X
10EE	E6 03	LDAB	3,X
10F0	EB 07	ADDB	7,X
10F2	A9 06	ADCA	6,X
10F4	B7 18 26	STAA	COR3
10F7	F7 18 27	STAB	COR4
10FA	7F 18 2A	CLR	COR9
10FD	7F 18 2B	CLR	COR10
1100	BD 13 85	JSR	ROT15
1103	FE 18 C0	LDX	I
1106	B6 18 24	LDAA	COR1
1109	A7 00	STAA	0,X
110B	B6 18 25	LDAA	COR2
110E	A7 01	STAA	1,X
1110	B6 18 26	LDAA	COR3
1113	A7 02	STAA	2,X
1115	B6 18 27	LDAA	COR4
1118	A7 03	STAA	3,X
111A	CE 18 C2	LDX	#BUT1R
111D	A6 00	LDAA	0,X
111F	E6 01	LDAB	1,X
1121	E0 05	SUBB	5,X
1123	A2 04	SBCA	4,X
1125	B7 18 24	STAA	COR1
1128	F7 18 25	STAB	COR2
112B	A6 02	LDAA	2,X
112D	E6 03	LDAB	3,X
112F	E0 07	SUBB	7,X
1131	A2 06	SBCA	6,X
1133	B7 18 26	STAA	COR3
1136	F7 18 27	STAB	COR4
1139	FE 18 BC	LDX	THETA
113C	FF 18 2A	STX	COR9
113F	BD 13 85	JSR	ROT15
1142	FE 18 CA	LDX	IP
1145	B6 18 24	LDAA	COR1
1148	A7 00	STAA	0,X
114A	B6 18 25	LDAA	COR2
114D	A7 01	STAA	1,X
114F	B6 18 26	LDAA	COR3
1152	A7 02	STAA	2,X
1154	B6 18 27	LDAA	COR4
1157	A7 03	STAA	3,X
1159	CE 18 B3	LDX	#N
115C	A6 02	LDAA	2,X
115E	E6 03	LDAB	3,X
1160	EB 0E	ADDB	14D,X
1162	A9 0D	ADCA	13D,X
1164	A7 0D	STAA	13D,X
1166	E7 0E	STAB	14D,X
1168	A6 02	LDAA	2,X
116A	E6 03	LDAB	3,X
116C	EB 1A	ADDB	26D,X

I=I+LE

# METHOD OF INVARIANT FOURIER SIGNATURES

116E	A9 19		ADCA	25D,X	
1170	A7 19		STAA	25D,X	II=II+LE
1172	E7 1A		STAB	26D,X	
1174	E0 01		SUBB	1,X	II-N
1176	A2 00		SBCA	0,X	
1178	2D 07		BLT	F7A	
117A	2E 08		BGT	F8	
117C	5D		TSTB		
117D	27 02		BEQ	F7A	
117F	20 03		BRA	F8	
1181	7E 10	9E F7A	JMP	F7	
1184	A6 09	F8	LDAA	9,X	
1186	E6 0A		LDAB	10D,X	
1188	EB 05		ADDB	5,X	
118A	A9 04		ADCA	4,X	
118C	A7 09		STAA	9,X	THETA=THETA+ANGLE
118E	E7 0A		STAB	10D,X	
1190	A6 0B		LDAA	11D,X	
1192	E6 0C		LDAB	12D,X	
1194	CB 04		ADDB	#4	
1196	89 00		ADCA	#0	
1198	A7 0B		STAA	11D,X	J=J+4
119A	E7 0C		STAB	12D,X	
119C	E0 08		SUBB	8,X	J-LE1
119E	A2 07		SBCA	7,X	
11A0	2D 07		BLT	F8A	
11A2	2E 08		BGT	F9	
11A4	5D		TSTB		
11A5	27 02		BEQ	F8A	
11A7	20 03		BRA	F9	
11A9	7E 10	85 F8A	JMP	F6	
11AC	6C 06	F7	INC	6,X	L=L+1
11AE	A6 06		LDAA	6,X	
11B0	B0 18	B1	SUBA	M	L-M
11B3	2E 03		BGT	F9A	
11B5	7E 10	52	JMP	F5	
11B8	A6 00	F9A	LDAA	0,X	BIT REVERSAL
11BA	E6 01		LDAB	1,X	
11BC	47		ASRA		
11BD	56		RORB		
11BE	A7 1B		STAA	27D,X	NV2=N/2
11C0	E7 1C		STAB	28D,X	
11C2	A6 00		LDAA	0,X	
11C4	E6 01		LDAB	1,X	
11C6	C0 04		SUBB	#4	
11C8	82 00		SBCA	#0	
11CA	A7 1D		STAA	29D,X	NM1=N-4
11CC	E7 1E		STAB	30D,X	
11CE	6F 21		CLR	33D,X	
11D0	86 04		LDAA	#4	
11D2	A7 22		STAA	34D,X	JJ=4
11D4	6F 19		CLR	25D,X	
11D6	A7 1A		STAA	26D,X	II=4
11D8	30		TSX		
11D9	08		INX		
11DA	08		INX		
11DB	FF 18 C0		STX	I	I=STKPTR+3

# METHOD OF INVARIANT FOURIER SIGNATURES

11DE	FF 18 BE	STX	J	J=I
11E1	CE 18 BA F10	LDX	#LE1	
11E4	A6 06	LDAA	6,X	
11E6	E6 07	LDAB	7,X	
11E8	E0 05	SUBB	5,X	I-J
11EA	A2 04	SBCA	4,X	
11EC	2C 4A	BGE	F11	
11EE	FE 18 BE	LDX	J	
11F1	A6 00	LDAA	0,X	
11F3	B7 18 C2	STAA	BUT1R	
11F6	A6 01	LDAA	1,X	
11F8	B7 18 C3	STAA	BUT1R+1	
11FB	A6 02	LDAA	2,X	
11FD	B7 18 C4	STAA	BUT1I	
1200	A6 03	LDAA	3,X	
1202	B7 18 C5	STAA	BUT1I+1	
1205	FE 18 C0	LDX	I	
1208	A6 00	LDAA	0,X	
120A	E6 01	LDAB	1,X	
120C	FE 18 BE	LDX	J	
120F	A7 00	STAA	0,X	
1211	E7 01	STAB	1,X	
1213	FE 18 C0	LDX	I	
1216	A6 02	LDAA	2,X	
1218	E6 03	LDAB	3,X	
121A	FE 18 BE	LDX	J	
121D	A7 02	STAA	2,X	
121F	E7 03	STAB	3,X	
1221	FE 18 C0	LDX	I	
1224	B6 18 C2	LDAA	BUT1R	
1227	A7 00	STAA	0,X	
1229	B6 18 C3	LDAA	BUT1R+1	
122C	A7 01	STAA	1,X	
122E	B6 18 C4	LDAA	BUT1I	
1231	A7 02	STAA	2,X	
1233	B6 18 C5	LDAA	BUT1I+1	
1236	A7 03	STAA	3,X	
1238	CE 18 BA F11	LDX	#LE1	
123B	A6 14	LDAA	20D,X	
123D	E6 15	LDAB	21D,X	
123F	A7 18	STAA	24D,X	K=NV2
1241	E7 19	STAB	25D,X	
1243	E0 1B F12	SUBB	27D,X	K-JJ
1245	A2 1A	SBCA	26D,X	
1247	2C 24	BGE	F13	
1249	A6 04	LDAA	4,X	
124B	E6 05	LDAB	5,X	
124D	E0 19	SUBB	25D,X	
124F	A2 1B	SBCA	24D,X	
1251	A7 04	STAA	4,X	J=J-K
1253	E7 05	STAB	5,X	
1255	A6 1A	LDAA	26D,X	
1257	E6 1B	LDAB	27D,X	
1259	E0 19	SUBB	25D,X	
125B	A2 1B	SBCA	24D,X	
125D	A7 1A	STAA	26D,X	JJ=JJ-K
125F	E7 1B	STAB	27D,X	

# METHOD OF INVARIANT FOURIER SIGNATURES

1261	A6 18		LDAA	24D,X	
1263	E6 19		LDAB	25D,X	
1265	47		ASRA		
1266	54		RORB		
1267	A7 18		STAA	24D,X	K=K/2
1269	E7 19		STAB	25D,X	
126B	20 D6		BRA	F12	
126D	A6 04	F13	LDAA	4,X	
126F	E6 05		LDAB	5,X	
1271	EB 19		ADDB	25D,X	
1273	A9 18		ADCA	24D,X	
1275	A7 04		STAA	4,X	J=J+K
1277	E7 05		STAB	5,X	
1279	A6 06		LDAA	6,X	
127B	E6 07		LDAB	7,X	
127D	CB 04		ADDB	#4	
127F	89 00		ADCA	#0	
1281	A7 06		STAA	6,X	I=I+4
1283	E7 07		STAB	7,X	
1285	A6 1A		LDAA	26D,X	
1287	E6 1B		LDAB	27D,X	
1289	EB 19		ADDB	25D,X	
128B	A9 18		ADCA	24D,X	
128D	A7 1A		STAA	26D,X	JJ=JJ+K
128F	E7 1B		STAB	27D,X	
1291	A6 12		LDAA	18D,X	
1293	E6 13		LDAB	19D,X	
1295	CB 04		ADDB	#4	
1297	89 00		ADCA	#0	
1299	A7 12		STAA	18D,X	II=II+4
129B	E7 13		STAB	19D,X	
129D	E0 17		SUBB	23D,X	II-NM1
129F	A2 16		SBCA	22D,X	
12A1	2D 06		BLT	F15	
12A3	2E 03		BGT	F14	
12A5	5D		TSTB		
12A6	27 01		BEQ	F15	
12A8	39	F14	RTS		
12A9	7E 11 E1	F15	JMP	F10	

# METHOD OF INVARIANT FOURIER SIGNATURES

			*	SUBROUTINE BSCALE
			*	BFP SCALING OF DATA IN STACK
			*	VECTOR CONSISTS OF N/2, 2 BYTE
			*	NUMBERS. ENTER WITH N IN X REG.
			*	RETURNS BLOCK SCALE FACTOR IN A REG.
12AC	FF 18 D6	BSCALE	STX	BSN
12AF	FF 18 DC		STX	BSN1
12B2	CE 00 00		LDX	#0
12B5	FF 18 D8		STX	POS
12B8	CE FF FF		LDX	FFFF
12BB	FF 18 DA		STX	NEG
12BE	30	BS1	TSX	
12BF	08		INX	
12C0	08		INX	
12C1	08	BS1A	INX	
12C2	08		INX	
12C3	E6 01		LDAB	1,X
12C5	A6 00		LDAA	0,X
12C7	2B 0E		BHI	BS2
12C9	BA 18 D8		ORAA	POS
12CC	FA 18 D9		ORAB	POS+1
12CF	B7 18 D8		STAA	POS
12D2	F7 18 D9		STAB	POS+1
12D5	20 0C		BRA	BS0
12D7	B4 18 DA	BS2	ANDA	NEG
12DA	F4 18 DB		ANDB	NEG+1
12DD	B7 18 DA		STAA	NEG
12E0	F7 18 DB		STAB	NEG+1
12E3	B6 18 D6	BS0	LDAA	BSN
12E6	F6 18 D7		LDAB	BSN+1
12E9	C0 02		SUBB	#2
12EB	82 00		SBCA	#0
12ED	B7 18 D6		STAA	BSN
12F0	F7 18 D7		STAB	BSN+1
12F3	2E CC		BGT	BS1A
12F5	5D		TSTB	
12F6	26 C9		BNE	BS1A
12F8	B6 18 DA		LDAA	NEG
12FB	F6 18 DB		LDAB	NEG+1
12FE	CE 00 00		LDX	#0
1301	58	BS4	ASLB	
1302	49		ROLA	
1303	24 03		BCC	BS5
1305	08		INX	
1306	20 F9		BRA	BS4
1308	09	BS5	DEX	
1309	09		DEX	
130A	FF 18 DA		STX	NEG
130D	B6 18 DB		LDAA	POS
1310	F6 18 D9		LDAB	POS+1
1313	CE 00 00		LDX	#0
1316	0D		SEC	
1317	59		ROLB	
1318	20 01		BRA	BS6A
131A	58	BS6	ASLB	
131B	49	BS6A	ROLA	
131C	25 03		BCS	BS7

# METHOD OF INVARIANT FOURIER SIGNATURES

131E	08			INX	
131F	20	F9		BRA	BS4
1321	09		BS7	DEX	
1322	09			DEX	
1323	FF	18	D8	STX	POS
1326	CE	18	D8	LDX	#POS
1329	E6	01		LDAB	1,X
132B	E0	03		SUBB	3,X
132D	2B	04		BMI	BS8
132F	A6	03		LDAA	3,X
1331	A7	01		STAA	1,X
1333	6D	01	BS8	TST	1,X
1335	26	02		BNE	BS9
1337	4F			CLRA	
1338	39			RTS	BLOCK S.F.=0
1339	30		BS9	TSX	
133A	08			INX	
133B	08			INX	
133C	08		BS9A	INX	SCALE DATA
133D	08			INX	
133E	B6	18	D9	LDAA	POS+1
1341	2B	10		BMI	BS11
1343	B7	18	D8	STAA	POS
1346	A6	00		LDAA	0,X
1348	E6	01		LDAB	1,X
134A	58		BS10	ASLB	
134B	49			ROLA	
134C	7A	18	D8	DEC	POS
134F	2E	F9		BGT	BS10
1351	20	06		BRA	BS12
1353	A6	00	BS11	LDAA	0,X
1355	E6	01		LDAB	1,X
1357	47			ASRA	
1358	56			RORB	
1359	A7	00	BS12	STAA	0,X
135B	E7	01		STAB	1,X
135D	B6	18	DC	LDAA	BSN1
1360	F6	18	DD	LDAB	BSN1+1
1363	C0	02		SUBB	#2
1365	82	00		SBCA	#0
1367	B7	18	DC	STAA	BSN1
136A	F7	18	DD	STAB	BSN1+1
136D	2E	CD		BGT	BS9A
136F	5D			TSTB	
1370	26	CA		BNE	BS9A
1372	B6	18	D9	LDAA	POS+1
1375	39			RTS	



# METHOD OF INVARIANT FOURIER SIGNATURES

		* * *		CORDIC ALGORITHM ITER ITERATIONS	
137A	86 0F	VECT15	LDAA	#15D	VECTOR: 15 ITERATIONS
137B	B7 18 0F		STAA	ITER	
137C	7F 18 2A		CLR	COR9	
137E	7F 18 2B	VECT	CLR	COR10	VECTOR: ITER ITERATIONS
1381	86 01		LDAA	#1	
1383	20 07		BRA	LCOR0	
1385	86 0F	ROT15	LDAA	#15D	ROTATION: 15 ITERATIONS
1387	B7 18 0F		STAA	ITER	
138A	86 80		LDAA	#80	
138B	CE 14 D9	LCOR0	LDX	#DLO	ROTATION: ITER ITERATIONS
138F	FF 18 0D		STX	COR19	
1392	CE 18 00		LDX	#COR5	
1395	A7 0C		STAA	#C,X	
1397	A6 2A		LDAA	#2A,X	
1399	A7 0A		STAA	#A,X	
139B	6F 0B		CLR	#B,X	
139D	A6 24		LDAA	#24,X	
139F	E6 25		LDAB	#25,X	
13A1	BD 14 91		JSR	SCALE	
13A4	A7 00		STAA	0,X	
13A6	A6 26		LDAA	#26,X	
13AB	E6 27		LDAB	#27,X	
13AA	BD 14 91		JSR	SCALE	
13AD	A7 01		STAA	1,X	
13AF	A6 00		LDAA	0,X	
13B1	A4 01		ANDA	1,X	
13B3	A7 01		STAA	1,X	
13B5	27 0C		BEQ	LCOR1	
13B7	A6 25		LDAA	#25,X	
13B9	A7 24		STAA	#24,X	
13BB	6F 25		CLR	#25,X	
13BD	A6 27		LDAA	#27,X	
13BF	A7 26		STAA	#26,X	
13C1	6F 27	LCOR1	CLR	#27,X	
13C3	6F 02		CLR	2,X	
13C5	6F 03		CLR	3,X	
13C7	6F 08		CLR	8,X	
13C9	6F 09		CLR	9,X	
13CB	6F 28		CLR	#28,X	
13CD	A6 24		LDAA	#24,X	
13CF	84 E0		ANDA	#E0	
13D1	2C 01		BGE	LCOR2	
13D3	43	LCOR2	COMA		
13D4	E6 26		LDAB	#26,X	
13D6	C4 E0		ANDB	#E0	
13D8	2C 01		BGE	LCOR3	
13DA	53		COMB		
13DB	E7 29		STAB	#29,X	
13DD	C6 01	LCOR3	LDAB	#1	
13DF	E7 00		STAB	0,X	
13E1	AA 29		ORAA	#29,X	
13E3	27 05		BEQ	LCOR3A	
13E5	BD 14 BC		JSR	DOWN2	
13E8	6A 00		DEC	0,X	

# METHOD OF INVARIANT FOURIER SIGNATURES

13EA	6F	04	LCOR3A	CLR	4,X	
13EC	A6	0F		LDAA	\$F,X	
13EE	A7	05		STAA	5,X	
13F0	CE	15 32		LDX	#CTABLE	
13F3	FF	18 06		STX	COR13	
13F6	FE	18 06	LCOR4	LDX	COR13	MAIN LOOP
13F9	A6	00		LDAA	0,X	
13FB	E6	01		LDAB	1,X	
13FD	08			INX		
13FE	08			INX		
13FF	FF	18 06		STX	COR13	
1402	FE	18 0D		LDX	COR19	
1405	08			INX		
1406	08			INX		
1407	FF	18 0D		STX	COR19	
140A	CE	18 00		LDX	#COR5	
140D	BD	14 AB		JSR	TSTCFL	
1410	2C	03		BGE	LCOR5	
1412	BD	14 A4		JSR	COMP	
1415	EB	2B	LCOR5	ADDB	\$2B,X	
1417	A9	2A		ADCA	\$2A,X	
1419	A7	2A		STAA	\$2A,X	
141B	E7	2B		STAB	\$2B,X	THETA DONE
141D	A6	26		LDAA	\$26,X	
141F	E6	27		LDAB	\$27,X	
1421	BD	14 AB		JSR	TSTCFL	
1424	2C	03		BGE	LCOR6	
1426	BD	14 A4		JSR	COMP	
1429	BD	14 D5	LCOR6	JSR	DOWN	
142C	CE	18 00		LDX	#COR5	
142F	EB	03		ADDB	3,X	
1431	A9	02		ADCA	2,X	
1433	A7	02		STAA	2,X	
1435	E7	03		STAB	3,X	X DONE
1437	A6	24		LDAA	\$24,X	
1439	E6	25		LDAB	\$25,X	
143B	BD	14 AB		JSR	TSTCFL	
143E	2D	03		BLT	LCOR7	
1440	BD	14 A4		JSR	COMP	
1443	BD	14 D5	LCOR7	JSR	DOWN	
1446	CE	18 00		LDX	#COR5	
1449	EB	09		ADDB	9,X	
144B	A9	08		ADCA	8,X	
144D	A7	26		STAA	\$26,X	
144F	E7	27		STAB	\$27,X	Y DONE
1451	A6	2A		LDAA	\$2A,X	
1453	A7	0A		STAA	\$A,X	
1455	6A	05		DEC	5,X	
1457	EE	02		LDX	2,X	
1459	FF	18 24		STX	COR1	
145C	FE	18 26		LDX	COR3	
145F	FF	18 08		STX	COR15	
1462	B6	18 05		LDAA	COR12	
1465	26	8F		BNE	LCOR4	
1467	CE	18 00		LDX	#COR5	
146A	6D	00		TST	0,X	
146C	27	03		BEQ	LCOR8	

# METHOD OF INVARIANT FOURIER SIGNATURES

146E	BD 14 BC		JSR	DOWN2
1471	6D 01	LCOR9	TST	1,X
1473	27 1B		BEQ	LCOR12
1475	A6 24		LDAA	\$24,X
1477	A7 25		STAA	\$25,X
1479	2A 06		BPL	LCOR9
147B	86 FF		LDAA	\$\$FF
147D	A7 24		STAA	\$24,X
147F	20 02		BRA	LCOR10
1481	6F 24	LCOR9	CLR	\$24,X
1483	A6 26	LCOR10	LDAA	\$26,X
1485	A7 27		STAA	\$27,X
1487	2A 05		BPL	LCOR11
1489	86 FF		LDAA	\$\$FF
148B	A7 26		STAA	\$26,X
148D	39		RTS	
148E	6F 26	LCOR11	CLR	\$26,X
1490	39	LCOR12	RTS	
* * * RETURNS RESCALE FACTOR IN A				
1491	4D	SCALE	TSTA	
1492	27 07		BEQ	LSC3
1494	43		COMA	
1495	27 03		BEQ	LSC2
1497	86 00	LSC1	LDAA	\$0
1499	39		RTS	
149A	50	LSC2	NEGB	
149B	C4 80	LSC3	ANDB	\$\$80
149D	27 02		BEQ	LSC4
149F	20 F6		BRA	LSC1
14A1	86 01	LSC4	LDAA	\$1
14A3	39		RTS	
* * * RETURNS 2'S COMP IN A,B				
14A4	50	COMP	NEGB	
14A5	25 02		BCS	LCP1
14A7	40		NEGA	
14A8	39		RTS	
14A9	43	LCP1	COMA	
14AA	39		RTS	
* * * RETURNS CONDITION CODE REGISTER STATE				
14AB	6D 0C	TSTCFL	TST	\$C,X
14AD	2E 0A		BGT	LT2
14AF	6D 0A		TST	\$A,X
14B1	2C 03		BGE	LT1
14B3	6D 0B		TST	\$B,X
14B5	39		RTS	
14B6	6D 0C	LT1	TST	\$C,X
14B8	39		RTS	
14B9	6D 26	LT2	TST	\$26,X
14BB	39		RTS	
* *				

# METHOD OF INVARIANT FOURIER SIGNATURES

14BC	A6	24	DOWN2	LDAA	\$24,X
14BE	E6	25		LDAB	\$25,X
14C0	47			ASRA	
14C1	56			RORB	
14C2	47			ASRA	
14C3	56			RORB	
14C4	A7	24		STAA	\$24,X
14C6	E7	25		STAB	\$25,X
14C8	A6	26		LDAA	\$26,X
14CA	E6	27		LDAB	\$27,X
14CC	47			ASRA	
14CD	56			RORB	
14CE	47			ASRA	
14CF	56			RORB	
14D0	A7	26		STAA	\$26,X
14D2	E7	27		STAB	\$27,X
14D4	39			RTS	
* * * RETURNS A,B SCALED DOWN 2**-(I-2) BITS					
14D5	EE	0D	DOWN	LDX	\$D,X
14D7	6E	00		JMP	0,X
14D9	20	2E	DL0	BRA	DL8
14DB	39			RTS	
14DC	39			RTS	
14DD	39			RTS	
14DE	39			RTS	
14DF	20	26		BRA	DL7
14E1	20	22		BRA	DL6
14E3	20	1E		BRA	DL5
14E5	20	1A		BRA	DL4
14E7	20	16		BRA	DL3
14E9	20	12		BRA	DL2
14EB	20	0E		BRA	DL1
14ED	20	1B		BRA	DL9
14EF	20	1C		BRA	DL11
14F1	20	1E		BRA	DL12
14F3	20	20		BRA	DL13
14F5	20	22		BRA	DL14
14F7	20	24		BRA	DL15
14F9	20	26		BRA	DL16
14FB	47		DL1	ASRA	
14FC	56			RORB	
14FD	47		DL2	ASRA	
14FE	56			RORB	
14FF	47		DL3	ASRA	
1500	56			RORB	
1501	47		DL4	ASRA	
1502	56			RORB	
1503	47		DL5	ASRA	
1504	56			RORB	
1505	47		DL6	ASRA	
1506	56			RORB	
1507	47		DL7	ASRA	
1508	56			RORB	
1509	39		DL8	RTS	

# METHOD OF INVARIANT FOURIER SIGNATURES

150A	8D 1E	DL9	BSR	SWITCH
150C	39	DL10	RTS	
150D	8D 1B	DL11	BSR	SWITCH
150F	20 17		BRA	DL22
1511	8D 17	DL12	BSR	SWITCH
1513	20 12		BRA	DL21
1515	8D 13	DL13	BSR	SWITCH
1517	20 0D		BRA	DL20
1519	8D 0F	DL14	BSR	SWITCH
151B	20 08		BRA	DL19
151D	8D 0B	DL15	BSR	SWITCH
151F	20 03		BRA	DL18
1521	8D 07	DL16	BSR	SWITCH
1523	57		ASRB	
1524	57	DL18	ASRB	
1525	57	DL19	ASRB	
1526	57	DL20	ASRB	
1527	57	DL21	ASRB	
1528	57	DL22	ASRB	
1529	39		RTS	

\*

152A	16	SWITCH	TAB	
152B	2A 03		BPL	SW1
152D	86 FF		LDAA	##FF
152F	39		RTS	
1530	4F	SW1	CLRA	
1531	39		RTS	

\*

1532		CTABLE	EQU	*
1532	40 00		FCB	\$40,0
1534	20 00		FCB	\$20,0
1536	12 E4		FCB	\$12,\$E4
1538	09 FB		FCB	9,\$FB
153A	05 11		FCB	5,\$11
153C	02 8B		FCB	2,\$8B
153E	01 46		FCB	1,\$46
1540	00 A3		FCB	0,\$A3
1542	00 51		FCB	0,\$51
1544	00 29		FCB	0,\$29
1546	00 14		FCB	0,\$14
1548	00 0A		FCB	0,\$A
154A	00 05		FCB	0,5
154C	00 03		FCB	0,3
154E	00 01		FCB	0,1
1550	00 01		FCB	0,1
1552			END	

STATEMENTS =762

FREE BYTES =16485

NO ERRORS DETECTED

# METHOD OF INVARIANT FOURIER SIGNATURES

1800		ORG	\$1800
1800	TEMP1	RMB	1
1801	TEMP2	RMB	1
1802	TEMP3	RMB	1
1803	TEMP4	RMB	1
1804	TEMP5	RMB	1
1805	TEMP6	RMB	1
1806	BCD1	RMB	1
1807	BCD2	RMB	1
1808	CTR	RMB	1
1809	CTR1	RMB	1
1824		ORG	\$1824
1824	COR1	RMB	1
1825	COR2	RMB	1
1826	COR3	RMB	1
1827	COR4	RMB	3
182A	COR9	RMB	1
182B	COR10	RMB	5
1830	UI1	RMB	1
1831	UI2	RMB	1
1832	VJ1	RMB	1
1833	VJ2	RMB	2
1835	DTIME	RMB	8
183D	NSAMP	RMB	1
183E	LOGS	RMB	16
184E	STACK1	RMB	1
184F	STACK2	RMB	1
1850	PUSHST	RMB	1
18A1		ORG	\$18A1
18A1	SIGNI	RMB	1
18A2	SIGNQ	RMB	1
18EC		ORG	\$18EC
18EC	I	RMB	1
18ED	P	RMB	2
18EF	THETA	RMB	4
18F3	PHI	RMB	2
18FF		ORG	\$18FF
18FF	INT	RMB	4
1903	S1	RMB	8
190B	S2	RMB	8
1913	S3	RMB	8
191B	AVG	RMB	6
8000	MATH	EQU	\$8000
330F	SYNDET	EQU	\$330F
33B8	MUXSEL	EQU	\$33B8
1376	VECT15	EQU	\$1376
2700		ORG	\$2700
	*	SUBROUTINE RCORR	
	*	CORRELATION OF TWO REAL SEQUENCES	
	*	CLEAR S1,S2,S3 FLOAT PT N8.	
	*	ACCUMULATE IN EXTERNAL LOOP; ENTER AT RCORR1	
	*	X POINTS TO CURRENT SAMPLE	
	*	ENTER AT RCORR2 TO COMPUTE $S1/(S2*S3)**0.5$	
2700	FF 18 00	RCORR1	STX TEMP1
2703	A6 00		LDAA 0,X
2705	E6 01		LDAB 1,X
2707	BD 28 7E		JSR PUSH82

# METHOD OF INVARIANT FOURIER SIGNATURES

270A	FE 18 00	LDX	TEMP1	
270D	A6 02	LDAA	2,X	
270F	E6 03	LDAB	3,X	
2711	BD 28 7E	JSR	PUSH82	
2714	86 01	LDAA	#1	
2716	BD 2A EB	JSR	MATH1	$X(I)*Y(I)$
2719	CE 19 03	LDX	#S1	
271C	BD 28 64	JSR	PUSH88	
271F	86 06	LDAA	#6	
2721	BD 80 00	JSR	MATH	$S1=S1+X(I)*Y(I)$
2724	CE 19 03	LDX	#S1	
2727	BD 28 4A	JSR	PULL8	
272A	FE 18 00	LDX	TEMP1	
272D	A6 00	LDAA	0,X	
272F	E6 01	LDAB	1,X	
2731	BD 28 7E	JSR	PUSH82	
2734	FE 18 00	LDX	TEMP1	
2737	A6 00	LDAA	0,X	
2739	E6 01	LDAB	1,X	
273B	BD 28 7E	JSR	PUSH82	
273E	86 01	LDAA	#1	
2740	BD 2A EB	JSR	MATH1	$X(I)*X(I)$
2743	CE 19 0B	LDX	#S2	
2746	BD 28 64	JSR	PUSH88	
2749	86 06	LDAA	#6	
274B	BD 80 00	JSR	MATH	$S2=S2+X(I)*X(I)$
274E	CE 19 0B	LDX	#S2	
2751	BD 28 4A	JSR	PULL8	
2754	FE 18 00	LDX	TEMP1	
2757	A6 02	LDAA	2,X	
2759	E6 03	LDAB	3,X	
275B	BD 28 7E	JSR	PUSH82	
275E	FE 18 00	LDX	TEMP1	
2761	A6 02	LDAA	2,X	
2763	E6 03	LDAB	3,X	
2765	BD 28 7E	JSR	PUSH82	
2768	86 01	LDAA	#1	
276A	BD 2A EB	JSR	MATH1	$Y(I)*Y(I)$
276D	CE 19 13	LDX	#S3	
2770	BD 28 64	JSR	PUSH88	
2773	86 06	LDAA	#6	
2775	BD 80 00	JSR	MATH	$S3=Y(I)*Y(I)$
2778	CE 19 13	LDX	#S3	
277B	BD 28 4A	JSR	PULL8	
277E	39	RTS		
277F	CE 19 0B	LDX	#S2	
2782	BD 28 64	JSR	PUSH88	
2785	CE 19 13	LDX	#S3	
2788	BD 28 64	JSR	PUSH88	
278B	86 01	LDAA	#1	
278D	BD 2A EB	JSR	MATH1	$S2*S3$
2790	CE 19 0B	LDX	#S2	
2793	BD 28 4A	JSR	PULL8	
2796	FE 19 0B	LDX	S2	COPY
2799	FF 19 13	STX	S3	
279C	FE 19 0D	LDX	S2+2	
279F	FF 19 15	STX	S3+2	

# METHOD OF INVARIANT FOURIER SIGNATURES

27A2	FE 19 0F		LDX	S2+4	
27A5	FF 19 17		STX	S3+4	
27A8	FE 19 11		LDX	S2+6	
27AB	FF 19 19		STX	S3+6	
27AE	B6 19 1A		LDA	S3+7	EXP
27B1	47		ASF		
27B2	B7 19 1A		STA	S3+7	S3=INITIAL ITERATE
27B5	86 0A		LDA	#10D	
27B7	B7 18 00		STA	TEMP1	
27BA	CE 19 08	RC1	LDY	#S2	
27BD	BD 28 64		JSF	PUSH88	
27C0	CE 19 13		LDY	#S3	
27C3	BD 28 64		JSF	PUSH88	
27C6	86 02		LDA	#2	
27C8	BD 80 00		JSF	MATH	X=S2/S3
27CB	CE 19 13		LDX	#S3	
27CE	BD 28 64		JSF	PUSH88	
27D1	86 06		LDA	#6	
27D3	BD 80 00		JSR	MATH	X=X+S3
27D6	30		TSX		
27D7	6A 07		DEC	7,X	X=0.5X
27D9	CE 19 13		LDY	#S3	
27DC	BD 28 4A		JSF	PULL8	
27DF	7A 18 00		DEC	TEMP1	
27E2	26 D6		BNE	RC1	
27E4	CE 19 03		LDX	#S1	SQUARE ROOT DONE
27E7	BD 28 64		JSF	PUSH88	
27EA	CE 19 13		LDY	#S3	
27ED	BD 28 64		JSF	PUSH88	
27F0	86 02		LDA	#2	
27F2	BD 80 00		JSF	MATH	
27F5	CE 19 03		LDY	#S1	
27F8	BD 28 4A		JSF	PULL8	S1=S1/(S2*S3)**0.5
27FB	F6 19 0A		LDA	S1+7	
27FE	2E 43		BGT	RC4	
2800	50		NE		
2801	CE 19 03		LDY	#S1	
2804	5D	RC2	TS		FLOAT TO FIX
2805	27 17		BEG	RC3	
2807	A6 00		LDA	0,X	
2809	47		ASH		
280A	A7 00		ST	0,X	
280C	A6 01		LDA	1,X	
280E	46		ROF		
280F	A7 01		ST	1,X	
2811	A6 02		LDA	2,X	
2813	46		ROF		
2814	A7 02		ST	2,X	
2816	A6 03		LDA	3,X	
2818	46		ROF		
2819	A7 03		ST	3,X	
281B	5A		DEC		
281C	20 E6		BR	RC2	
281E	CE 19 03	RC3	LDY	#S1	
2821	4F		CLF		
2822	BD 28 C9		JSF	PUSH44	
2825	86 DC		LDA	#DC	



# METHOD OF INVARIANT FOURIER SIGNATURES

2827	36		PSHA		
2828	86	46	LDAA	#46	
282A	36		PSHA		
282B	86	03	LDAA	#03	
282D	36		PSHA		
282E	4F		CLRA		
282F	36		PSHA		SF=65536*32768/10000
2830	86	09	LDAA	#9	
2832	BD	80 00	JSR	MATH	SCALE FOR BCD
2835	CE	19 03	LDX	#S1	
2838	4F		CLRA		
2839	BD	28 A3	JSR	PULL4	
283C	FE	19 05	LDX	S1+2	
283F	FF	18 02	STX	TEMP3	
2842	39		RTS		
2843	CE	27 0F	LDX	#270F	EXP>0
2846	FF	18 02	STX	TEMP3	
2849	39		RTS		

\*  
\* SUBROUTINE PULL8  
\* PULLS FLOATING PT N  
\* X POINTS TO MSB

284A	33		PULL8	PULB	
284B	F7	18 4E		STAB	STACK1
284E	33			PULB	
284F	F7	18 4F		STAB	STACK2
2852	C6	08		LDAB	#8
2854	32		P8	PULA	
2855	A7	00		STAA	0,X
2857	08			INX	
2858	5A			DECB	
2859	26	F9		BNE	P8
285B	F6	18 4F		LDAB	STACK2
285E	37			PSHB	
285F	F6	18 4E		LDAB	STACK1
2862	37			PSHB	
2863	39			RTS	

\*  
\* SUBROUTINE PUSH88  
\* PUSHES N FLOATING PT N ON STACK  
\* X POINTS TO MSB

2864	33		PUSH88	PULB	
2865	F7	18 4E		STAB	STACK1
2868	33			PULB	
2869	F7	18 4F		STAB	STACK2
286C	C6	08		LDAB	#8
286E	A6	07	P88	LDAA	7,X
2870	36			PSHA	
2871	09			DEX	
2872	5A			DECB	
2873	26	F9		BNE	P88
2875	F6	18 4F		LDAB	STACK2
2878	37			PSHB	
2879	F6	18 4E		LDAB	STACK1
287C	37			PSHB	
287D	39			RTS	

# METHOD OF INVARIANT FOURIER SIGNATURES

```

*
* SUBROUTINE PUSH82
* PUSHES 2 BYTE NUMBER ON STACK
* CONVERTS TO FLOAT
*
287E F7 18 50 PUSH82 STAB PUSHST
2881 33 PULB
2882 F7 18 4E STAB STACK1
2885 33 PULB
2886 F7 18 4F STAB STACK2
2889 5F CLRB
288A 37 PSHB
288B 37 PSHB
288C 37 PSHB
288D 37 PSHB
288E 37 PSHB
288F 37 PSHB
2890 F6 18 50 LDAB PUSHST
2893 37 PSHB
2894 36 PSHA
2895 86 07 LDAA #7
2897 8D 80 00 JSR MATH NFPN
289A F6 18 4F LDAB STACK2
289D 37 PSHB
289E F6 18 4E LDAB STACK1
28A1 37 PSHB
28A2 39 RTS

```

```

*
* SUBROUTINE PULL4
* PULLS 4 BYTES OFF STACK
* STORES MSB IN X+A
*
28A3 33 PULL4 PULB
28A4 F7 18 4E STAB STACK1
28A7 33 PULB
28AB F7 18 4F STAB STACK2
28AB 4D TSTA
28AC 27 06 BEQ LP42
28AE 08 LP41 INX
28AF 4A DECA
28B0 27 02 BEQ LP42
28B2 20 FA BRA LP41
28B4 32 LP42 PULA
28B5 A7 00 STAA 0,X
28B7 32 PULA
28B8 A7 01 STAA 1,X
28BA 32 PULA
28BB A7 02 STAA 2,X
28BD 32 PULA
28BE A7 03 STAA 3,X
28C0 F6 18 4F LDAB STACK2
28C3 37 PSHB
28C4 F6 18 4E LDAB STACK1
28C7 37 PSHB
28C8 39 RTS

```

```

*
* SUBROUTINE PUSH44
* PUSHES 4 BYTE NUMBER ONTO STACK
* MSB IN X+A
*

```

# METHOD OF INVARIANT FOURIER SIGNATURES

28C9	33	PUSH44	PULB	
28CA	F7 18 4E		STAB	STACK1
28CD	33		PULB	
28CE	F7 18 4F		STAB	STACK2
28D1	4D		TSTA	
28D2	27 06		BEQ	LP442
28D4	08	LP441	INX	
28D5	4A		DECA	
28D6	27 02		BEQ	LP442
28D8	20 FA		BRA	LP441
28DA	A6 03	LP442	LDAA	3,X
28DC	36		PSHA	
28DD	A6 02		LDAA	2,X
28DF	36		PSHA	
28E0	A6 01		LDAA	1,X
28E2	36		PSHA	
28E3	A6 00		LDAA	0,X
28E5	36		PSHA	
28E6	F6 18 4F		LDAB	STACK2
28E9	37		PSHB	
28EA	F6 18 4E		LDAB	STACK1
28ED	37		PSHB	
28EE	39		RTS	
* * * * * SUBROUTINE PUSH41 PUSHES 4 BYTE NUMBER ONTO STACK 3 MS BYTES ARE SIGN BITS ENTER WITH LS BYTE IN A REG				
28EF	33	PUSH41	PULB	
28F0	F7 18 4E		STAB	STACK1
28F3	33		PULB	
28F4	F7 18 4F		STAB	STACK2
28F7	36		PSHA	
28F8	48		ASLA	
28F9	24 04		BCC	LP411
28FB	86 FF		LDAA	##FF
28FD	20 01		BRA	LP412
28FF	4F	LP411	CLRA	POSITIVE
2900	36	LP412	PSHA	
2901	36		PSHA	
2902	36		PSHA	
2903	F6 18 4F		LDAB	STACK2
2906	37		PSHB	
2907	F6 18 4E		LDAB	STACK1
290A	37		PSHB	
290B	39		RTS	
* * * * * SUBROUTINE PUSH42 PUSHES 4 BYTE NUMBER ONTO STACK 2 MS BYTER ARE SIGN BITS ENTER WITH LS BYTE IN B, NEXT LS BYTER IN A				
290C	F7 18 50	PUSH42	STAB	PUSHST
290F	33		PULB	
2910	F7 18 4E		STAB	STACK1
2913	33		PULB	
2914	F7 18 4F		STAB	STACK2
2917	F6 18 50		LDAB	PUSHST

# METHOD OF INVARIANT FOURIER SIGNATURES

291A	37		PSHB		
291B	36		PSHA		
291C	48		ASLA		
291D	24	04	BCC	LP421	
291F	86	FF	LDAA	\$FF	NEGATIVE
2921	20	01	BRA	LP422	
2923	4F		CLRA		POSITIVE
2924	36		PSHA		
2925	36		PSHA		
2926	F6	18 4F	LDAB	STACK2	
2929	37		PSHB		
292A	F6	18 4E	LDAB	STACK1	
292D	37		PSHB		
292E	39		RTS		

* * * SUBROUTINE TRAPIN TRAPEZOIDAL INTEGRATION					
292F	F6	18 ED	TRAPIN	LDAB	P
2932	B6	19 02	T1	LDAA	INT+3
2935	BB	18 25		ADDA	COR2
2938	B7	19 02		STAA	INT+3
293B	B6	19 01		LDAA	INT+2
293E	B9	18 24		ADCA	COR1
2941	B7	19 01		STAA	INT+2
2944	B6	19 00		LDAA	INT+1
2947	B9	00		ADCA	\$0
2949	B7	19 00		STAA	INT+1
294C	B6	18 FF		LDAA	INT
294F	B9	00		ADCA	\$0
2951	B7	18 FF		STAA	INT
2954	5D			TSTB	
2955	27	03		BEG	T2
2957	5A			DECB	
2958	20	D8		BRA	T1
295A	86	01	T2	LDAA	\$1
295C	B7	18 ED		STAA	P
295F	39			RTS	

* * * SUBROUTINE TRAPFX FINAL SAMPLE: TRAP INT					
2960	86	02	TRAPFX	LDAA	\$2
2962	BD	29 99		JSR	DELAY1
2965	BD	2A 4C		JSR	RDDEFT
2968	BD	13 76		JSR	VECT15
296B	7F	18 ED		CLR	P
296E	BD	29 2F		JSR	TRAPIN
2971	CE	18 FF		LDX	\$INT
2974	A6	21		LDAA	\$21,X
2976	AB	03		ADDA	3,X
2978	A7	21		STAA	\$21,X
297A	A6	20		LDAA	\$20,X
297C	A9	02		ADCA	2,X
297E	A7	20		STAA	\$20,X
2980	A6	1F		LDAA	\$1F,X
2982	A9	01		ADCA	1,X
2984	A7	1F		STAA	\$1F,X
2986	A6	1E		LDAA	\$1E,X

# METHOD OF INVARIANT FOURIER SIGNATURES

2988	A9	00		ADCA	0,X
298A	A7	1E		STAA	\$1E,X
298C	A6	1D		LDAA	\$1D,X
298E	89	00		ADCA	\$0
2990	A7	1D		STAA	\$1D,X
2992	A6	1C		LDAA	\$1C,X
2994	89	00		ADCA	\$0
2996	A7	1C		STAA	\$1C,X
2998	39			RTS	

\*  
\* SUBROUTINE DELAY1  
\* 9.996MS\*DTIME DELAY

2999	B6	18	35	DELAY1	LDAA	DTIME
299C	4D			D1	TSTA	
299D	27	05			BEQ	D2
299F	8D	04			BSR	DELAY
29A1	4A				DECA	
29A2	20	F8			BRA	D1
29A4	39			D2	RTS	

\*  
\* SUBROUTINE DELAY

29A5	CE	02	CA	DELAY	LDX	\$2CA
29A8	09			LDY1	DEX	
29A9	27	02			BEQ	LDY2
29AB	20	F8			BRA	LDY1
29AD	39			LDY2	RTS	

\*  
\* SUBROUTINE BI\$BCD  
\* POSITIVE BINARY IN TEMP3,TEMP4  
\* RETURNS BCD IN BCD1,BCD2

29AE	7F	18	06	BI\$BCD	CLR	BCD1
29B1	7F	18	07		CLR	BCD2
29B4	86	10			LDAA	\$16D
29B6	B7	18	09		STAA	CTR1
29B9	BF	18	04		STS	TEMP5
29BC	8E	18	01		LDS	\$TEMP3-1
29BF	86	08		LBIB1	LDAA	\$8D
29C1	B7	18	08		STAA	CTR
29C4	33				PULB	

29C5	58			LBIB2	ASLB	
29C6	24	12			BCC	LBIB3
29C8	86	01			LDAA	\$1
29CA	BB	18	07		ADDA	BCD2
29CD	19				DAA	
29CE	B7	18	07		STAA	BCD2
29D1	B6	18	06		LDAA	BCD1
29D4	89	00			ADCA	\$0
29D6	19				DAA	

29D7	B7	18	06		STAA	BCD1
29DA	7A	18	09	LBIB3	DEC	CTR1
29DD	27	18			BEQ	LBIB4
29DF	B6	18	07		LDAA	BCD2
29E2	BB	18	07		ADDA	BCD2
29E5	19				DAA	
29E6	B7	18	07		STAA	BCD2
29E9	B6	18	06		LDAA	BCD1
29EC	B9	18	06		ADCA	BCD1

# METHOD OF INVARIANT FOURIER SIGNATURES

29EF	19			DAA	
29F0	B7	18	06	STAA	BCD1
29F3	7A	18	08	DEC	CTR
29F6	27	C7		BEQ	LBIB1
29F8	20	CB		BRA	LBIB2
29FA	BE	18	04	LDS	TEMP5
29FD	39			RTS	

\*

\*

## SUBROUTINE TUNE

29FE	B6	18	30	TUNE	LDAA	UI1
2A01	B7	EE	21		STAA	\$EE21
2A04	B6	18	31		LDAA	UI2
2A07	B7	EE	20		STAA	\$EE20
2A0A	B6	18	32		LDAA	VJ1
2A0D	B7	EE	11		STAA	\$EE11
2A10	B6	18	33		LDAA	VJ2
2A13	B7	EE	10		STAA	\$EE10
2A16	39				RTS	

\*

\*

## SUBROUTINES IDEX

2A17	FF	18	00	IDEX	STX	TEMP1
2A1A	FB	18	01		ADDB	TEMP2
2A1D	B9	18	00		ADCA	TEMP1
2A20	B7	18	00		STAA	TEMP1
2A23	F7	18	01		STAB	TEMP2
2A26	FE	18	00		LDX	TEMP1
2A29	39				RTS	

\*

\*

## SUBROUTINE WRITE

2A2A	A6	00		WRITE	LDAA	0,X
2A2C	44				LSRA	
2A2D	44				LSRA	
2A2E	44				LSRA	
2A2F	44				LSRA	
2A30	8A	30			ORAA	\$30
2A32	BD	FD	80		JSR	\$FD80
2A35	A6	00			LDAA	0,X
2A37	84	0F			ANDA	\$0F
2A39	8A	30			ORAA	\$30
2A3B	BD	FD	80		JSR	\$FD80
2A3E	39				RTS	

\*

\*

## SUBROUTINE PASC

2A3F	5D			PASC	TSTB	
2A40	27	09			BEQ	PAS
2A42	A6	00			LDAA	0,X
2A44	BD	FD	80		JSR	\$FD80
2A47	08				INX	
2A48	5A				DECB	
2A49	20	F4			BRA	PASC
2A4B	39				RTS	

\*

\*

## SUBROUTINE RDDEFT

2A4C	B6	18	3D	RDDEFT	LDAA	NSAMP
2A4F	B7	18	08		STAA	CTR
2A52	CE	00	00		LDX	#0
2A55	FF	18	24		STX	COR1

# METHOD OF INVARIANT FOURIER SIGNATURES

2A5B	FF	18	26		STX	COR3
2A5B	FF	18	28		STX	COR4+1
2A5E	7D	18	08	RD1	TST	CTR
2A61	27	2D			BEQ	RD2
2A63	8D	63			BSR	READSN
2A65	CE	18	00		LDX	\$1800
2A68	A6	26			LDAA	\$26,X
2A6A	AB	01			ADDA	1,X
2A6C	A7	26			STAA	\$26,X
2A6E	A6	25			LDAA	\$25,X
2A70	A9	00			ADCA	0,X
2A72	A7	25			STAA	\$25,X
2A74	A6	24			LDAA	\$24,X
2A76	A9	A1			ADCA	\$A1,X
2A78	A7	24			STAA	\$24,X
2A7A	A6	29			LDAA	\$29,X
2A7C	AB	03			ADDA	3,X
2A7E	A7	29			STAA	\$29,X
2A80	A6	28			LDAA	\$28,X
2A82	A9	02			ADCA	2,X
2A84	A7	28			STAA	\$28,X
2A86	A6	27			LDAA	\$27,X
2A88	A9	A2			ADCA	\$A2,X
2A8A	A7	27			STAA	\$27,X
2A8C	6A	08			DEC	8,X
2A8E	20	CE			BRA	RD1
2A90	A6	3E	RD2		LDAA	\$3E,X
2A92	A7	08			STAA	8,X
2A94	6D	08	RD3		TST	8,X
2A96	27	23			BEQ	RD4
2A98	A6	24			LDAA	\$24,X
2A9A	47				ASRA	
2A9B	A7	24			STAA	\$24,X
2A9D	A6	25			LDAA	\$25,X
2A9F	46				RORA	
2AA0	A7	25			STAA	\$25,X
2AA2	A6	26			LDAA	\$26,X
2AA4	46				RORA	
2AA5	A7	26			STAA	\$26,X
2AA7	A6	27			LDAA	\$27,X
2AA9	47				ASRA	
2AAA	A7	27			STAA	\$27,X
2AAC	A6	28			LDAA	\$28,X
2AAE	46				RORA	
2AAF	A7	28			STAA	\$28,X
2AB1	A6	29			LDAA	\$29,X
2AB3	46				RORA	
2AB4	A7	29			STAA	\$29,X
2AB6	7A	18	08		DEC	CTR
2AB9	20	D9			BRA	RD3
2ABB	FE	18	25	RD4	LDX	COR2
2ABE	FF	18	24		STX	COR1
2AC1	FE	18	28		LDX	COR4+1
2AC4	FF	18	26		STX	COR3
2AC7	39				RTS	

\*  
\*

SUBROUTINE READSN

# METHOD OF INVARIANT FOURIER SIGNATURES

2AC8	84	03	READSN	LDAA	#3
2ACA	B7	33	B8	STAA	MUXSEL
2ACD	CE	18	00	LDX	#TEMP1
2AD0	BD	33	0F	JSR	SYNDET
2AD3	7D	18	00	TST	TEMP1
2AD6	8D	0C		BSR	SGNSET
2AD8	B7	18	A1	STAA	SIGNI
2ADB	7D	18	02	TST	TEMP3
2ADE	8D	04		BSR	SGNSET
2AE0	B7	18	A2	STAA	SIGNQ
2AE3	39			RTS	
*					
2AE4	2A	03	SGNSET	BPL	SETPOS
2AE6	86	FF		LDAA	#FF
2AE8	39			RTS	
2AE9	4F		SETPOS	CLRA	
2AEA	39			RTS	
*					
2AEB	33		MATH1	PULB	
2AEC	F7	18	4E	STAB	STACK1
2AEF	33			PULB	
2AF0	F7	18	4F	STAB	STACK2
2AF3	30			TSX	
2AF4	E6	0F		LDAB	15D,X
2AF6	C1	80		CMPI	#80
2AF8	27	05		BEQ	M11
2AFA	BD	80	00	JSR	MATH
2AFD	20	06		BRA	M13
2AFF	86	08	M11	LDAA	#8
2B01	33		M12	PULB	
2B02	4A			DECA	
2B03	26	FC		BNE	M12
2B05	F6	18	4F	LDAB	STACK2
2B08	37			PSHB	
2B09	F6	18	4E	LDAB	STACK1
2B0C	37			PSHB	
2B0D	39			RTS	
*					
2B0E	BD	FD	A6	PRINT	JSR
2B11	CE	18	EC	LDX	\$FDA6
2B14	C6	01		LDAB	#1
2B16	BD	FD	74	JSR	\$FD74
2B19	86	20		LDAA	#20
2B1B	BD	FD	80	JSR	\$FD80
2B1E	CE	18	30	LDX	#UI1
2B21	BD	2A	2A	JSR	WRITE
2B24	CE	18	31	LDX	#UI2
2B27	BD	2A	2A	JSR	WRITE
2B2A	86	20		LDAA	#20
2B2C	BD	FD	80	JSR	\$FD80
2B2F	CE	18	32	LDX	#VJ1
2B32	BD	2A	2A	JSR	WRITE
2B35	CE	18	33	LDX	#VJ2
2B38	BD	2A	2A	JSR	WRITE
2B3B	86	20		LDAA	#20
2B3D	BD	FD	80	JSR	\$FD80
2B40	CE	18	EF	LDX	#THETA



# METHOD OF INVARIANT FOURIER SIGNATURES

2B43	C6 02	LDAB	#2
2B45	BD FD 74	JSR	\$FD74
2B48	86 20	LDAA	##20
2B4A	BD FD 80	JSR	\$FD80
2B4D	CE 18 F3	LDX	#PHI
2B50	C6 02	LDAB	#2
2B52	BD FD 74	JSR	\$FD74
2B55	39	RTS	
2B56		END	

STATEMENTS =637

FREE BYTES =17972

NO ERRORS DETECTED

ASSEMBLE

## APPENDIX B - Method of Invariant Moments Assembly Code Listing

This Appendix consists of a listing of the assembly language program which computes invariant Fourier signatures. This program was written to run on the Deft Laboratories' microprocessor-based test bed. All addresses and opcodes are hexadecimal. In the operand column of the statements the following symbols are used:

\$	Hexadecimal Prefix
%	Binary Prefix
H	Hexadecimal Postfix
D	Decimal Postfix
B	Binary Postfix
#	Denotes Immediate Addressing Mode

The entry address for this program is \$2000.

# METHOD OF INVARIANT MOMENTS

1802		ORG	\$1802
1802	TEMP3	RMB	1
1803	TEMP4	RMB	3
1806	BCD1	RMB	1
1807	BCD2	RMB	1
182A		ORG	\$182A
182A	COR9	RMB	1
182B	COR10	RMB	5
1830	UI1	RMB	1
1831	UI2	RMB	1
1832	VJ1	RMB	1
1833	VJ2	RMB	2
1835	DTIME	RMB	1
1836	I	RMB	1
1837	J	RMB	1
1838	K	RMB	1
1839	XST	RMB	2
183B	VUJ1	RMB	1
183C	VUJ2	RMB	1
183D	NSAMP	RMB	1
183E	LOGS	RMB	1
183F	DX	RMB	2
1841	DY	RMB	2
1843	DDY	RMB	2
1845	PO	RMB	2
1847	PX	RMB	2
1849	PY	RMB	2
184B	SIGN	RMB	3
184E	STACK1	RMB	1
184F	STACK2	RMB	1
1850	PUSHST	RMB	1
1851	LIMIT	RMB	1
18E0		ORG	\$18E0
18E0	IPX	RMB	2
18E2	IPY	RMB	2
18E4	X	RMB	8
18EC	Y	RMB	8
1B50		ORG	\$1B50
1B50	M00	RMB	\$200
8000	MATH	EQU	\$8000
29EC	MATH1	EQU	\$29EC
2AF0	BI\$BCD	EQU	\$2AF0
2800	FL\$BCD	EQU	\$2800
281F	FL2	EQU	\$281F
2961	SAMP	EQU	\$2961
2976	SAMPLE	EQU	\$2976
2A0F	FZERO	EQU	\$2A0F
29D7	INCX	EQU	\$29D7
2872	ACCUM	EQU	\$2872
2B5A	PUSH88	EQU	\$2B5A
2B40	PULL8	EQU	\$2B40
2A2A	SQROOT	EQU	\$2A2A
2C17	PASC	EQU	\$2C17
2C02	WRITE	EQU	\$2C02
2CC2	C01	EQU	\$2CC2
2D12	C23	EQU	\$2D12
2D62	C45	EQU	\$2D62

# METHOD OF INVARIANT MOMENTS

2DB2				C67	EQU	\$2DB2	
2000					ORG	\$2000	
2000	BD	FD	A6	CALIB	JSR	\$FDA6	
2003	CE	24	97		LDX	\$LINE1	
2006	C6	24			LDAB	\$36D	
2008	BD	2C	17		JSR	PASC	
2008	BD	FD	36	CB1	JSR	\$FD36	INCH
200E	81	43			CMFA	\$843	
2010	26	F9			BNE	CB1	
2012	FE	25	77		LDX	XZERO	MEASURE P0,PX,PY
2015	FF	18	30		STX	UI1	
2018	FE	25	79		LDX	YZERO	
2018	FF	18	32		STX	UJ1	
201E	86	80			LDAA	\$880	
2020	B7	18	3D		STAA	NSAMP	
2023	86	07			LDAA	\$7	
2025	B7	18	3E		STAA	LOGS	
2028	86	02			LDAA	\$2	
202A	B7	18	35		STAA	DTIME	
202D	BD	29	61		JSR	SAMP	
2030	B7	18	45		STAA	P0	
2033	F7	18	46		STAB	P0+1	
2036	CE	99	93		LDX	\$99993	
2039	FF	18	3F		STX	DX	
203C	CE	00	07		LDX	\$7	
203F	FF	18	41		STX	DY	
2042	BD	29	D7		JSR	INCX	
2045	BD	29	61		JSR	SAMP	
2048	F0	18	46		SUBB	P0+1	
2048	B2	18	45		SBCA	P0	
204E	B7	18	47		STAA	PX	
2051	F7	18	48		STAB	PX+1	
2054	FE	25	77		LDX	XZERO	
2057	FF	18	30		STX	UI1	
205A	BD	24	48		JSR	INCY	
205D	BD	29	61		JSR	SAMP	
2060	F0	18	46		SUBB	P0+1	
2063	B2	18	45		SBCA	P0	
2066	B7	18	49		STAA	PY	
2069	F7	18	4A		STAB	PY+1	
206C	BD	FD	A6	FEAT#2	JSR	\$FDA6	ENTER TO SKIP CALIBRATION
206F	CE	24	BB		LDX	\$LINE2	
2072	C6	26			LDAB	\$38D	
2074	BD	2C	17		JSR	PASC	
2077	BD	FD	A6		JSR	\$FDA6	
207A	BD	FD	A6	FE1	JSR	\$FDA6	
207D	CE	24	E1		LDX	\$LINE3	
2080	C6	15			LDAB	\$21D	
2082	BD	2C	17		JSR	PASC	
2085	BD	FD	36		JSR	\$FD36	
2088	84	0F			ANDA	\$80F	
208A	B7	18	3E		STAA	LOGS	
208D	C6	01			LDAB	\$1	
208F	4A			FE2	DECA		
2090	2D	03			BLT	FE3	
2092	58				ASLB		
2093	20	FA			BRA	FE2	

# METHOD OF INVARIANT MOMENTS

2095	F7 18 3D	FE3	STAB	NSAMP	
2098	BD FD A6		JSR	\$FDA6	
209B	CE 25 71		LDX	\$LINE11	
209E	C6 06		LDAB	\$6	
20A0	BD 2C 17		JSR	PASC	
20A3	BD FD 36		JSR	\$FD36	
20A6	84 0F		ANDA	\$0F	
20A8	B7 18 51		STAA	LIMIT	
20AB	BD FD A6	FE3A	JSR	\$FDA6	
20AE	CE 24 F6		LDX	\$LINE4	
20B1	C6 21		LDAB	\$33D	
20B3	BD 2C 17		JSR	PASC	
20B6	BD FD 36	FE4	JSR	\$FD36	
20B9	81 4D		CMPA	\$4D	
20BB	26 F9		BNE	FE4	
20BD	FE 25 79		LDX	YZERO	
20C0	FF 18 32		STX	VJ1	
20C3	FF 18 3B		STX	VUJ1	
20C6	CE 99 93		LDX	\$9993	
20C9	FF 18 3F		STX	DX	
20CC	FF 18 43		STX	DDY	
20CF	CE 00 07		LDX	\$7	
20D2	FF 18 41		STX	DY	
20D5	86 02		LDAA	\$2	
20D7	B7 18 35		STAA	DTIME	
20DA	CE 00 00		LDX	\$0	
20DD	FF 18 E2		STX	IPY	
20E0	CE 18 50		LDX	\$M00	
20E3	86 40		LDAA	\$64	
20E5	BD 2A 0F	FE5	JSR	FZERO	INITIALIZE M(P,Q)
20E8	4A		DECA		
20E9	26 FA		BNE	FE5	
20EB	7F 18 37		CLR	J	
20EE	B6 18 37	FE6	LDAA	J	WY LOOP
20F1	27 04		BEQ	FE7	
20F3	84 03		ANDA	\$3	
20F5	27 4F		BEQ	FE13A	
20F7	FE 25 77	FE7	LDX	XZERO	INITIALIZE WX LOOP
20FA	FF 18 30		STX	UI1	
20FD	CE 00 00		LDX	\$0	
2100	FF 18 E0		STX	IPX	
2103	7F 18 36		CLR	I	
2106	B6 18 36	FE8	LDAA	I	WX LOOP
2109	27 07		BEQ	FE9	
210B	84 03		ANDA	\$3	
210D	27 39		BEQ	FE13B	
210F	44		LSRA		
2110	25 0D		BCS	FE10	
2112	BD 29 76	FE9	JSR	SAMPLE	X**P EVEN
2115	B6 18 37		LDAA	J	
2118	27 12		BEQ	FE11	
211A	44		LSRA		
211B	25 31		BCS	FE14	
211D	20 0D		BRA	FE11	
211F	BD 29 76	FE10	JSR	SAMPLE	X**P ODD

# METHOD OF INVARIANT MOMENTS

2122	B6	18	37		LDAA	J	
2125	27	43			BEQ	FE17	
2127	44				LSRA		
2128	25	5D			BCS	FE20	
212A	20	3E			BRA	FE17	
212C	CE	18	50	FE11	LDX	#M00	***P EVEN, Y**Q EVEN
212F	7F	18	38		CLR	K	
2132	B6	18	38	FE12	LDAA	K	
2135	5F				CLRB		
2136	84	09			ANDA	#9	
2138	26	01			BNE	FE13	
213A	5C				INCB		
213B	4F			FE13	CLRA		
213C	BD	28	72		JSR	ACCUM	
213F	8C	1D	50		CPX	#M00+512D	
2142	26	EE			BNE	FE12	
2144	20	5C			BRA	FE23	
2146	20	6A		FE13A	BRA	FE24	
2148	20	58		FE13B	BRA	FE23	
214A	20	A2		FE13C	BRA	FE6	
214C	20	88		FE13D	BRA	FE8	
214E	CE	18	50	FE14	LDX	#M00	***P EVEN, Y**Q ODD
2151	7F	18	38		CLR	K	
2154	B6	18	38	FE15	LDAA	K	
2157	84	09			ANDA	#9	
2159	5F				CLRB		
215A	4A				DECA		
215B	26	01			BNE	FE16	
215D	5C				INCB		
215E	86	01		FE16	LDAA	#1	
2160	BD	28	72		JSR	ACCUM	
2163	8C	1D	50		CPX	#M00+512D	
2166	26	EC			BNE	FE15	
2168	20	38			BRA	FE23	
216A	CE	18	50	FE17	LDX	#M00	***P ODD, Y**Q EVEN
216D	7F	18	38		CLR	K	
2170	B6	18	38	FE18	LDAA	K	
2173	84	09			ANDA	#9	
2175	5F				CLRB		
2176	81	08			CMPA	#8	
2178	26	01			BNE	FE19	
217A	5C				INCB		
217B	86	02		FE19	LDAA	#2	
217D	BD	28	72		JSR	ACCUM	
2180	8C	1D	50		CPX	#M00+512D	
2183	26	EB			BNE	FE18	
2185	20	1B			BRA	FE23	
2187	CE	18	50	FE20	LDX	#M00	***P ODD, Y**Q ODD
218A	7F	18	38		CLR	K	
218D	B6	18	38	FE21	LDAA	K	
2190	84	09			ANDA	#9	
2192	5F				CLRB		
2193	81	09			CMPA	#9	
2195	26	01			BNE	FE22	
2197	5C				INCB		
2198	86	03		FE22	LDAA	#3	
219A	BD	28	72		JSR	ACCUM	

# METHOD OF INVARIANT MOMENTS

219D	8C 1D 50		CPX	#M00+512D	
21A0	26 EB		BNE	FE21	
21A2	7C 18 36	FE23	INC	I	
21A5	BD 24 71		JSR	INCXP	
21A8	BD 29 D7		JSR	INCX	
21AB	B6 18 36		LDAA	I	
21AE	81 28		CMPA	#40D	
21B0	26 9A		BNE	FE13D	
21B2	7C 18 37	FE24	INC	J	
21B5	BD 24 84		JSR	INCYP	
21B8	BD 24 48		JSR	INCY	
21BB	B6 18 37		LDAA	J	
21BE	81 28		CMPA	#40D	
21C0	26 88		BNE	FE13C	
21C2	CE 1B 58		LDX	#M00+8	NORMALIZE MOMENTS
21C5	FF 18 39		STX	XST	
21C8	FE 18 39	FE25	LDX	XST	
21CB	BD 2B 5A		JSR	PUSH88	
21CE	CE 1B 50		LDX	#M00	
21D1	BD 2B 5A		JSR	PUSH88	
21D4	B6 02		LDAA	#2	
21D6	BD 80 00		JSR	MATH	M(P,Q)/M(0,0)
21D9	FE 18 39		LDX	XST	
21DC	BD 2B 40		JSR	PULLB	
21DF	FF 18 39		STX	XST	
21E2	8C 1D 50		CPX	#M00+512D	
21E5	26 E1		BNE	FE25	
21E7	CE 1B 60		LDX	#M00+16D	M(0,2)
21EA	BD 2B 5A		JSR	PUSH88	
21ED	CE 1B D0		LDX	#M00+128D	M(2,0)
21F0	BD 2B 5A		JSR	PUSH88	
21F3	B6 06		LDAA	#6	
21F5	BD 80 00		JSR	MATH	
21F8	CE 1B 58		LDX	#M00+8	M(0,1)
21FB	BD 2B 5A		JSR	PUSH88	
21FE	CE 1B 58		LDX	#M00+8	
2201	BD 2B 5A		JSR	PUSH88	
2204	B6 01		LDAA	#1	
2206	BD 29 EC		JSR	MATH1	
2209	B6 05		LDAA	#5	
220B	BD 80 00		JSR	MATH	
220E	CE 1B 90		LDX	#M00+64D	M(1,0)
2211	BD 2B 5A		JSR	PUSH88	
2214	CE 1B 90		LDX	#M00+64D	
2217	BD 2B 5A		JSR	PUSH88	
221A	B6 01		LDAA	#1	
221C	BD 29 EC		JSR	MATH1	
221F	B6 05		LDAA	#5	
2221	BD 80 00		JSR	MATH	
2224	CE 1B 50		LDX	#M00	
2227	BD 2B 5A		JSR	PUSH88	
222A	B6 02		LDAA	#2	
222C	BD 80 00		JSR	MATH	
222F	CE 1B E4		LDX	#X	
2232	BD 2B 40		JSR	PULLB	X MOMENT INVARIANT
2235	CE 1B 98		LDX	#M00+72D	M(1,1)
2238	BD 2B 5A		JSR	PUSH88	

# METHOD OF INVARIANT MOMENTS

223B	CE 1B 5B	LDX	#M00+8	M(0,1)
223E	BD 2B 5A	JSR	PUSH88	
2241	CE 1B 90	LDX	#M00+64D	M(1,0)
2244	BD 2B 5A	JSR	PUSH88	
2247	86 01	LDAA	#1	
2249	BD 29 EC	JSR	MATH1	
224C	86 05	LDAA	#5	
224E	BD 80 00	JSR	MATH	
2251	CE 1B EC	LDX	#Y	
2254	BD 2B 40	JSR	PULL8	
2257	CE 1B EC	LDX	#Y	
225A	BD 2B 5A	JSR	PUSH88	
225D	CE 1B EC	LDX	#Y	
2260	BD 2B 5A	JSR	PUSH88	
2263	86 01	LDAA	#1	
2265	BD 29 EC	JSR	MATH1	
2268	30	TSX		
2269	6C 07	INC	7,X	
226B	6C 07	INC	7,X	
226D	CE 1B 5B	LDX	#M00+8	M(0,1)
2270	BD 2B 5A	JSR	PUSH88	
2273	CE 1B 5B	LDX	#M00+8	
2276	BD 2B 5A	JSR	PUSH88	
2279	86 01	LDAA	#1	
227B	BD 29 EC	JSR	MATH1	
227E	CE 1B 90	LDX	#M00+64D	M(1,0)
2281	BD 2B 5A	JSR	PUSH88	
2284	CE 1B 90	LDX	#M00+64D	
2287	BD 2B 5A	JSR	PUSH88	
228A	86 01	LDAA	#1	
228C	BD 29 EC	JSR	MATH1	
228F	86 05	LDAA	#5	
2291	BD 80 00	JSR	MATH	
2294	CE 1B D0	LDX	#M00+128D	M(2,0)
2297	BD 2B 5A	JSR	PUSH88	
229A	86 06	LDAA	#6	
229C	BD 80 00	JSR	MATH	
229F	CE 1B 60	LDX	#M00+16D	M(0,2)
22A2	BD 2B 5A	JSR	PUSH88	
22A5	86 05	LDAA	#5	
22A7	BD 80 00	JSR	MATH	
22AA	CE 1B EC	LDX	#Y	
22AD	BD 2B 40	JSR	PULL8	
22B0	CE 1B EC	LDX	#Y	
22B3	BD 2B 5A	JSR	PUSH88	
22B6	CE 1B EC	LDX	#Y	
22B9	BD 2B 5A	JSR	PUSH88	
22BC	86 01	LDAA	#1	
22BE	BD 29 EC	JSR	MATH1	
22C1	86 06	LDAA	#6	
22C3	BD 80 00	JSR	MATH	Y#2
22C6	BD 2A 2A	JSR	SQR00T	
22C9	CE 1B 50	LDX	#M00	
22CC	BD 2B 5A	JSR	PUSH88	
22CF	86 02	LDAA	#2	
22D1	BD 80 00	JSR	MATH	
22D4	CE 1B EC	LDX	#Y	



# METHOD OF INVARIANT MOMENTS

22D7	BD	2B	40		JSR	PULLB	Y MOMENT INVARIANT
22DA	CE	1B	50		LDX	\$M00	
22DD	86	40			LDAA	\$140	
22DF	A7	00			STAA	0,X	
22E1	6F	01			CLR	1,X	
22E3	6F	02			CLR	2,X	
22E5	6F	03			CLR	3,X	
22E7	6F	04			CLR	4,X	
22E9	6F	05			CLR	5,X	
22EB	6F	06			CLR	6,X	
22ED	86	01			LDAA	\$1	
22EF	A7	07			STAA	7,X	
22F1	BD	FD	A6		JSR	\$FDA6	
22F4	CE	25	17		LDX	\$LINE5	PRINT MOMENTS
22F7	C6	06			LDAB	\$6	
22F9	BD	2C	17		JSR	PASC	
22FC	BD	FD	A6		JSR	\$FDA6	
22FF	CE	25	1D		LDX	\$LINE6	
2302	C6	3E			LDAB	\$62D	
2304	BD	2C	17		JSR	PASC	
2307	7F	18	36		CLR	I	
230A	CE	1B	50		LDX	\$M00	
230D	FF	18	39		STX	XST	
2310	BD	FD	A6	FE26	JSR	\$FDA6	
2313	B6	18	36		LDAA	I	
2316	8A	30			ORAA	\$130	
2318	BD	FD	80		JSR	\$FD80	
231B	7F	18	37		CLR	J	
231E	B6	18	37	FE27	LDAA	J	
2321	B1	18	51		CMPA	LIMIT	
2324	2A	25			BPL	FE28	
2326	CE	25	1D		LDX	\$LINE6	
2329	C6	02			LDAB	\$2	
232B	BD	2C	17		JSR	PASC	
232E	FE	18	39		LDX	XST	
2331	BD	2B	00		JSR	FL\$BCD	
2334	B6	18	4B		LDAA	SIGN	
2337	BD	FD	80		JSR	\$FD80	
233A	86	2E			LDAA	\$12E	
233C	BD	FD	80		JSR	\$FD80	
233F	CE	1B	06		LDX	\$BCD1	
2342	BD	2C	02		JSR	WRITE	
2345	CE	1B	07		LDX	\$BCD2	
2348	BD	2C	02		JSR	WRITE	
234B	FE	18	39	FE28	LDX	XST	
234E	0B				INX		
234F	0B				INX		
2350	0B				INX		
2351	0B				INX		
2352	0B				INX		
2353	0B				INX		
2354	0B				INX		
2355	0B				INX		
2356	FF	18	39		STX	XST	
2359	7C	18	37		INC	J	
235C	B6	18	37		LDAA	J	
235F	B1	0B			CMPA	\$B	

# METHOD OF INVARIANT MOMENTS

2361	26 BB	BNE	FE27
2363	7C 18 36	INC	I
2366	B6 18 36	LDAA	I
2369	B1 18 51	CHPA	LIMIT
236C	2B A2	BMI	FE26
236E	BD FD A6	JSR	\$FDA6
2371	BD FD A6	JSR	\$FDA6
2374	CE 25 5B	LDX	\$LINE7
2377	C6 03	LDAB	\$3
2379	BD 2C 17	JSR	PASC
237C	CE 18 E4	LDX	\$X
237F	BD 28 1F	JSR	FL2
2382	B6 18 4B	LDAA	SIGN
2385	BD FD 80	JSR	\$FD80
2388	86 2E	LDAA	\$2E
238A	BD FD 80	JSR	\$FD80
238D	CE 18 06	LDX	\$BCD1
2390	BD 2C 02	JSR	WRITE
2393	CE 18 07	LDX	\$BCD2
2396	BD 2C 02	JSR	WRITE
2399	CE 25 5E	LDX	\$LINE8
239C	C6 04	LDAB	\$4
239E	BD 2C 17	JSR	PASC
23A1	7D 18 EB	TST	X+7
23A4	2A 11	BPL	FE30
23A6	4F	CLRA	
23A7	5F	CLRB	
23A8	F0 18 EB	SUBB	X+7
23AB	82 FF	SBCA	\$FF
23AD	B7 18 02	STAA	TEMP3
23B0	F7 18 03	STAB	TEMP4
23B3	86 2D	LDAA	\$2D
23B5	20 0B	BRA	FE31
23B7	B6 18 EB	LDAA	X+7
23BA	B7 18 03	STAA	TEMP4
23BD	7F 18 02	CLR	TEMP3
23C0	86 2B	LDAA	\$2B
23C2	BD FD 80	JSR	\$FD80
23C5	BD 2A F0	JSR	BI\$BCD
23C8	CE 18 07	LDX	\$BCD2
23CB	BD 2C 02	JSR	WRITE
23CE	CE 25 62	LDX	\$LINE9
23D1	C6 05	LDAB	\$5
23D3	BD 2C 17	JSR	PASC
23D6	CE 18 EC	LDX	\$Y
23D9	BD 28 1F	JSR	FL2
23DC	B6 18 4B	LDAA	SIGN
23DF	BD FD 80	JSR	\$FD80
23E2	86 2E	LDAA	\$2E
23E4	BD FD 80	JSR	\$FD80
23E7	CE 18 06	LDX	\$BCD1
23EA	BD 2C 02	JSR	WRITE
23ED	CE 18 07	LDX	\$BCD2
23F0	BD 2C 02	JSR	WRITE
23F3	CE 25 5E	LDX	\$LINE8
23F6	C6 04	LDAB	\$4
23F8	BD 2C 17	JSR	PASC

# METHOD OF INVARIANT MOMENTS

23FB	7D 18 F3		TST	Y+7	
23FE	2A 11		BPL	FE34	
2400	4F		CLRA		
2401	5F		CLRB		
2402	F0 18 F3		SUBB	Y+7	
2405	B2 FF		SBCA	##FF	
2407	B7 18 02		STAA	TEMP3	
240A	F7 18 03		STAB	TEMP4	
240D	B6 2D		LDAA	##2D	
240F	20 0B		BRA	FE35	
2411	B6 18 F3	FE34	LDAA	Y+7	
2414	B7 18 03		STAA	TEMP4	
2417	7F 18 02		CLR	TEMP3	
241A	B6 2B		LDAA	##2B	
241C	BD FD 80	FE35	JSR	\$FD80	
241F	BD 2A F0		JSR	BI\$BCD	
2422	CE 18 07		LDX	\$BCD2	
2425	BD 2C 02		JSR	WRITE	
2428	BD FD A6		JSR	\$FDA6	
242B	BD FD A6		JSR	\$FDA6	
242E	B6 07		LDAA	##07	
2430	BD FD 80		JSR	\$FD80	
2433	CE 25 67		LDX	\$LINE10	
2436	C6 0A		LDAB	\$10D	
2438	BD 2C 17		JSR	PASC	
243B	BD FD 36		JSR	\$FD36	
243E	81 59		CMPA	##59	
2440	26 03		BNE	FE36	
2442	7E 20 AB		JMP	FE3A	
2445	7E FE 2D	FE36	JMP	\$FE2D	FANTOM
* *					
2448	B6 18 33	INCY	LDAA	VJ2	
244B	BB 18 42		ADDA	DY+1	
244E	19		DAA		
244F	B7 18 33		STAA	VJ2	
2452	B6 18 32		LDAA	VJ1	
2455	B9 18 41		ADCA	DY	
2458	19		DAA		
2459	B7 18 32		STAA	VJ1	
245C	B6 18 3C		LDAA	VVJ2	
245F	BB 18 44		ADDA	DDY+1	
2462	19		DAA		
2463	B7 18 3C		STAA	VVJ2	
2466	B6 18 3B		LDAA	VVJ1	
2469	B9 18 43		ADCA	DDY	
246C	19		DAA		
246D	B7 18 3B		STAA	VVJ1	
2470	39		RTS		
* *					
2471	B6 18 E0	INCP	LDAA	IPX	
2474	F6 18 E1		LDAB	IPX+1	
2477	F8 18 48		ADDB	PX+1	
247A	B9 18 47		ADCA	PX	
247D	B7 18 E0		STAA	IPX	
2480	F7 18 E1		STAB	IPX+1	

# METHOD OF INVARIANT MOMENTS

2483 39 RTS

\*  
\*

2484	B6 18 E2	INCYP	LDAA	IPY
2487	F6 18 E3		LDAB	IPY+1
248A	FB 18 4A		ADDB	PY+1
248D	B9 18 49		ADCA	PY
2490	B7 18 E2		STAA	IPY
2493	F7 18 E3		STAB	IPY+1
2496	39		RTS	

\*  
\*

2497		LINE1	EQU	*
2497	50 4F 53		FCC	'POSITION UNIFORM IMAGE.'
24AE	20 54 4B		FCC	' THEN TYPE C '
24BB		LINE2	EQU	*
24BB	43 4F 4D		FCC	'COMPUTES M(P,Q) MOMENTS'
24D2	20 46 52		FCC	' FROM TRANSFORM'
24E1		LINE3	EQU	*
24E1	53 41 4D		FCC	'SAMPLE AVERAGE IS 2**'
24F6		LINE4	EQU	*
24F6	50 4F 53		FCC	'POSITION TEST IMAGE.'
250A	20 54 4B		FCC	' THEN TYPE M '
2517		LINE5	EQU	*
2517	4D 28 50		FCC	'M(P,Q)'
251D		LINE6	EQU	*
251D	20 20 20		FCC	' 0 1 2'
2533	20 20 20		FCC	' 3 4 5 6 7'
255B		LINE7	EQU	*
255B	5B 3D 20		FCC	'X= '
255E		LINE8	EQU	*
255E	20 32 2A		FCC	' 2**'
2562		LINE9	EQU	*
2562	20 20 59		FCC	' Y= '
2567		LINE10	EQU	*
2567	43 4F 4E		FCC	'CONTINUE? '
2571		LINE11	EQU	*
2571	4C 49 4D		FCC	'LIMIT= '
2577		XZERO	EQU	*
2577	3B 57		FCB	\$3B,\$57
2579		YZERO	EQU	*
2579	34 9B		FCB	\$34,\$9B
257B			END	

STATEMENTS =557

FREE BYTES =18705

NO ERRORS DETECTED

MBLE

# METHOD OF INVARIANT MOMENTS

1800		ORG	\$1800
1800	TEMP1	RMB	1
1801	TEMP2	RMB	1
1802	TEMP3	RMB	1
1803	TEMP4	RMB	1
1804	TEMP5	RMB	1
1805	TEMP6	RMB	1
1806	BCD1	RMB	1
1807	BCD2	RMB	1
1808	CTR	RMB	1
1809	CTR1	RMB	1
1824		ORG	\$1824
1824	COR1	RMB	1
1825	COR2	RMB	1
1826	COR3	RMB	1
1827	COR4	RMB	3
182A	COR9	RMB	1
182B	COR10	RMB	5
1830	UI1	RMB	1
1831	UI2	RMB	1
1832	VJ1	RMB	1
1833	VJ2	RMB	2
1835	DTIME	RMB	1
1836	I	RMB	1
1837	J	RMB	1
1838	K	RMB	1
1839	XST	RMB	2
183B	VVJ1	RMB	1
183C	VVJ2	RMB	1
183D	NSAMP	RMB	1
183E	LOGS	RMB	1
183F	DX	RMB	2
1841	DY	RMB	2
1843	DDY	RMB	2
1845	PO	RMB	2
1847	PX	RMB	2
1849	PY	RMB	2
184B	SIGN	RMB	3
184E	STACK1	RMB	1
184F	STACK2	RMB	1
1850	PUSHST	RMB	1
1851	LIMIT	RMB	1
18A1		ORG	\$18A1
18A1	SIGNI	RMB	1
18A2	SIGNQ	RMB	1
18E0		ORG	\$18E0
18E0	IPX	RMB	2
18E2	IPY	RMB	2
18E4	X	RMB	8
18EC	Y	RMB	8
18F4	REAL1	RMB	2
18F6	REAL2	RMB	2
18F8	IMAG1	RMB	2
18FA	IMAG2	RMB	2
8000	MATH	EQU	\$8000

# METHOD OF INVARIANT MOMENTS

1385		ROT15	EQU	\$1385
1376		VECT15	EQU	\$1376
330F		SYNDET	EQU	\$330F
33B8		HUXSEL	EQU	\$33B8
2800			ORG	\$2800
		*	CONVERTS FLPN TO BCD	
		*	X POINTS TO FLPN	
		*	NORMAL ENTRY REQUIRES MAG<1. IF NOT, BCD SET TO 9999	
2800	E6 07	FL\$BCD	LDAB	7,X NORMAL ENTRY
2802	2E 57		BGT	FL5
2804	50		NEGB	
2805	5D	FL1	TSTB	
2806	27 17		BEQ	FL2
2808	A6 00		LDAA	0,X
280A	47		ASRA	
280B	A7 00		STAA	0,X
280D	A6 01		LDAA	1,X
280F	46		RORA	
2810	A7 01		STAA	1,X
2812	A6 02		LDAA	2,X
2814	46		RORA	
2815	A7 02		STAA	2,X
2817	A6 03		LDAA	3,X
2819	46		RORA	
281A	A7 03		STAA	3,X
281C	5A		DECB	
281D	20 E6		BRA	FL1
281F	6D 00	FL2	TST	0,X FRACTION ONLY
2821	2A 14		BPL	FL3
2823	4F		CLRA	
2824	BD 2B E5		JSR	PUSH41
2827	4F		CLRA	
2828	BD 2B BF		JSR	PUSH44
282B	86 0C		LDAA	\$12D
282D	BD 80 00		JSR	MATH
2830	86 2D		LDAA	\$2D
2832	B7 18 4B		STAA	SIGN
2835	20 09		BRA	FL4
2837	4F	FL3	CLRA	
2838	BD 2B BF		JSR	PUSH44
283B	86 2B		LDAA	\$2B
283D	B7 18 4B		STAA	SIGN
2840	86 DC	FL4	LDAA	\$DC
2842	36		PSHA	
2843	86 46		LDAA	\$46
2845	36		PSHA	
2846	86 03		LDAA	\$03
2848	36		PSHA	
2849	4F		CLRA	
284A	36		PSHA	
284B	86 09		LDAA	\$9
284D	BD 80 00		JSR	MATH
2850	CE 18 00		LDX	\$TEMP1
2853	4F		CLRA	
2854	BD 2B 99		JSR	PULL4
2857	BD 2A F0		JSR	BI\$BCD

# METHOD OF INVARIANT MOMENTS

285A	39			RTS	
285B	6D 00	FL5		TST	0,X
285D	2A 07			BPL	FL6
285F	86 2D			LDAA	##2D
2861	B7 18 4B			STAA	SIGN
2864	20 05			BRA	FL7
2866	86 2B	FL6		LDAA	##2B
2868	B7 18 4B			STAA	SIGN
286B	CE 99 99	FL7		LDX	##9999
286E	FF 18 06			STX	BCD1
2871	39			RTS	
* * *					
2872	5D	ACCUM		TSTB	
2873	26 03			BNE	AC0
2875	7E 29 55	AC0A		JMP	AC13
2878	F6 18 38	AC0		LDAB	K
287B	C4 07			ANDB	#7
287D	F1 18 51			CMPB	LIMIT
2880	2A F3			BPL	AC0A
2882	F6 18 38			LDAB	K
2885	C4 38			ANDB	##38
2887	57			ASRB	
2888	57			ASRB	
2889	57			ASRB	
288A	F1 18 51			CMPB	LIMIT
288D	2A E6			BPL	AC0A
288F	FF 18 39			STX	XST
2892	16			TAB	
2893	27 14			BEQ	AC1
2895	5A			DECB	
2896	27 1F			BEQ	AC2
2898	5A			DECB	
2899	27 2A			BEQ	AC3
289B	B6 18 F6			LDAA	REAL2
289E	F6 18 F7			LDAB	REAL2+1
28A1	F0 18 F5			SUBB	REAL1+1
28A4	B2 18 F4			SBCA	REAL1
28A7	20 2B			BRA	AC4
28A9	B6 18 F4	AC1		LDAA	REAL1
28AC	F6 18 F5			LDAB	REAL1+1
28AF	F8 18 F7			ADDB	REAL2+1
28B2	B9 18 F6			ADCA	REAL2
28B5	20 1A			BRA	AC4
28B7	B6 18 F8	AC2		LDAA	IMAG1
28BA	F6 18 F9			LDAB	IMAG1+1
28BD	F0 18 FB			SUBB	IMAG2+1
28C0	B2 18 FA			SBCA	IMAG2
28C3	20 0C			BRA	AC4
28C5	B6 18 FB	AC3		LDAA	IMAG1
28C8	F6 18 F9			LDAB	IMAG1+1
28CB	FB 18 FB			ADDB	IMAG2+1
28CE	B9 18 FA			ADCA	IMAG2
28D1	BD 2B 74	AC4		JSR	PUSH82
28D4	B6 18 38			LDAA	K

# METHOD OF INVARIANT MOMENTS

28D7	B4	07		ANDA	#7	
28D9	47			ASRA		
28DA	27	0B		BEQ	AC5	
28DC	4A			DECA		
28DD	27	0D		BEQ	AC6	
28DF	4A			DECA		
28E0	27	0F		BEQ	AC7	
28E2	CE	2D	B2	LDX	#C67	
28E5	20	0D		BRA	AC8	
28E7	CE	2C	C2 AC5	LDX	#C01	
28EA	20	0B		BRA	AC8	
28EC	CE	2D	12 AC6	LDX	#C23	
28EF	20	03		BRA	AC8	
28F1	CE	2D	62 AC7	LDX	#C45	
28F4	4F		AC8	CLRA		
28F5	F6	18	37	LDAB	J	
28F8	58			ASLB		
28F9	49			ROLA		
28FA	BD	2A	C8	JSR	IDEX	
28FD	A6	00		LDAA	0,X	
28FF	E6	01		LDAB	1,X	
2901	BD	2B	74	JSR	PUSH82	CJ(Q)
2904	86	01		LDAA	#1	
2906	BD	29	EC	JSR	MATH1	
2909	B6	18	38	LDAA	K	
290C	84	38		ANDA	##38	
290E	47			ASRA		
290F	47			ASRA		
2910	47			ASRA		
2911	47			ASRA		
2912	27	0B		BEQ	AC9	
2914	4A			DECA		
2915	27	0D		BEQ	AC10	
2917	4A			DECA		
2918	27	0F		BEQ	AC11	
291A	CE	2D	B2	LDX	#C67	
291D	20	0D		BRA	AC12	
291F	CE	2C	C2 AC9	LDX	#C01	
2922	20	0B		BRA	AC12	
2924	CE	2D	12 AC10	LDX	#C23	
2927	20	03		BRA	AC12	
2929	CE	2D	62 AC11	LDX	#C45	
292C	4F		AC12	CLRA		
292D	F6	18	36	LDAB	I	
2930	58			ASLB		
2931	49			ROLA		
2932	BD	2A	C8	JSR	IDEX	
2935	A6	00		LDAA	0,X	
2937	E6	01		LDAB	1,X	
2939	BD	2B	74	JSR	PUSH82	CI(P)
293C	86	01		LDAA	#1	
293E	BD	29	EC	JSR	MATH1	
2941	FE	18	39	LDX	XST	
2944	BD	2B	5A	JSR	PUSH88	M(P,Q)
2947	86	06		LDAA	#6	
2949	BD	80	00	JSR	MATH	
294C	FE	18	39	LDX	XST	



AD-A110 322

DEFT LABS INC EAST SYRACUSE NY

F/B 20/1

ACOUSTO-OPTIC TECHNOLOGY FOR TOPOGRAPHIC FEATURE EXTRACTION AND--ETC(U)

MAR 81 A L MOYER

DAAK70-79-C-0160

UNCLASSIFIED

0102-A002

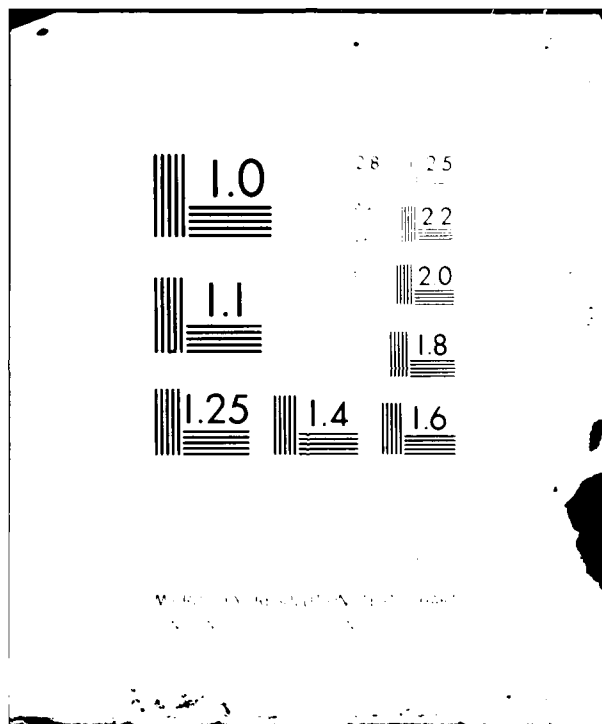
ETL-0256

NL

3 OF 3

AD-A  
110 322

END  
DATE  
FILMED  
03-82  
DTIC



# METHOD OF INVARIANT MOMENTS

294F	BD	2B	40	JSR	PULL8
2952	FE	18	39	LDX	XST
2955	08		AC13	INX	
2956	08			INX	
2957	08			INX	
2958	08			INX	
2959	08			INX	
295A	08			INX	
295B	08			INX	
295C	08			INX	
295D	7C	18	38	INC	K
2960	39			RTS	

\*  
\*  
\*

2961	BD	2A	96	SAMP	JSR	TUNE
2964	BD	2A	DB		JSR	DELAY1
2967	BD	2C	24		JSR	RDDEFT
296A	BD	13	76		JSR	VECT15
296D	4F				CLRA	
296E	5F				CLRB	
296F	F0	18	2B		SUBB	COR10
2972	B2	18	2A		SBCA	COR9
2975	39				RTS	

\*  
\*  
\*

2976	BD	2A	96	SAMPLE	JSR	TUNE	1ST QUAD
2979	BD	2A	DB		JSR	DELAY1	
297C	BD	2C	24		JSR	RDDEFT	
297F	B4	18	45		LDAA	P0	
2982	F6	18	46		LDAB	P0+1	
2985	FB	18	E1		ADDB	IPX+1	
2988	B9	18	E0		ADCA	IPX	
298B	FB	18	E3		ADDB	IPY+1	
298E	B9	18	E2		ADCA	IPY	
2991	B7	18	2A		STAA	COR9	
2994	F7	18	2B		STAB	COR10	
2997	BD	13	85		JSR	ROT15	

299A	FE	18	24		LDX	COR1	
299D	FF	18	F4		STX	REAL1	
29A0	FE	18	26		LDX	COR3	
29A3	FF	18	F8		STX	IMAG1	
29A6	BD	2A	AF		JSR	TUNE1	4TH QUAD
29A9	BD	2A	DB		JSR	DELAY1	
29AC	BD	2C	24		JSR	RDDEFT	
29AF	B4	18	45		LDAA	P0	
29B2	F6	18	46		LDAB	P0+1	
29B5	FB	18	E1		ADDB	IPX+1	
29B8	B9	18	E0		ADCA	IPX	
29BB	F0	18	E3		SUBB	IPY+1	
29BE	B2	18	E2		SBCA	IPY	
29C1	B7	18	2A		STAA	COR9	
29C4	F7	18	2B		STAB	COR10	
29C7	BD	13	85		JSR	ROT15	
29CA	FE	18	24		LDX	COR1	
29CD	FF	18	F4		STX	REAL2	

# METHOD OF INVARIANT MOMENTS

29D0	FE 18 26		LDX	COR3
29D3	FF 18 FA		STX	IMAG2
29D6	39		RTS	

\*  
\*  
\*

29D7	B6 18 31	INCX	LDAA	UI2
29DA	BB 18 40		ADDA	DX+1
29DD	19		DAA	
29DE	B7 18 31		STAA	UI2
29E1	B6 18 30		LDAA	UI1
29E4	B9 18 3F		ADCA	DX
29E7	19		DAA	
29E8	B7 18 30		STAA	UI1
29EB	39		RTS	

\*  
\*  
\*

29EC	33	MATH1	PULB	
29ED	F7 18 4E		STAB	STACK1
29F0	33		PULB	
29F1	F7 18 4F		STAB	STACK2
29F4	30		TSX	
29F5	E6 0F		LDAB	15D,X
29F7	C1 80		CHPB	##80
29F9	27 05		BEQ	M11
29FB	BD 80 00		JSR	MATH
29FE	20 06		BRA	M13
2A00	86 08	M11	LDAA	#8
2A02	33	M12	PULB	
2A03	4A		DECA	
2A04	26 FC		BNE	M12
2A06	F6 18 4F	M13	LDAB	STACK2
2A09	37		PSHB	
2A0A	F6 18 4E		LDAB	STACK1
2A0D	37		PSHB	
2A0E	39		RTS	

\*  
\*

2A0F	6F 00	F75R0	CLR	0,X
2A11	6F 01		CLR	1,X
2A13	6F 02		CLR	2,X
2A15	6F 03		CLR	3,X
2A17	6F 04		CLR	4,X
2A19	6F 05		CLR	5,X
2A1B	6F 06		CLR	6,X
2A1D	C6 80		LDAB	##80
2A1F	E7 07		STAB	7,X
2A21	08		INX	
2A22	08		INX	
2A23	08		INX	
2A24	08		INX	
2A25	08		INX	
2A26	08		INX	
2A27	08		INX	
2A28	08		INX	
2A29	39		RTS	

## METHOD OF INVARIANT MOMENTS

		* * *		SQUARE ROOT OF FLPTN IN STACK	
2A2A	33	SQROOT	PULB		
2A2B	F7 18 4E		STAB	STACK1	
2A2E	33		PULB		
2A2F	F7 18 4F		STAB	STACK2	
2A32	30		TSX		
2A33	C6 18		LDAB	#24D	
2A35	09	SQ1	DEX		
2A36	5A		DECB		
2A37	26 FC		BNE	SQ1	
2A39	FF 18 39		STX	XST	
2A3C	C6 08		LDAB	#8D	
2A3E	A6 1F	SQ2	LDAA	31D,X	
2A40	36		PSHA		
2A41	09		DEX		
2A42	5A		DECB		
2A43	26 F9		BNE	SQ2	
2A45	A6 1F		LDAA	31D,X	
2A47	47		ASRA		
2A48	A7 1F		STAA	31D,X	
2A4A	86 0A		LDAA	#10D	
2A4C	B7 18 08		STAA	CTR	
2A4F	FE 18 39	SQ3	LDX	XST	MAIN LOOP
2A52	C6 10		LDAB	#16D	
2A54	A6 1F	SQ4	LDAA	31D,X	
2A56	36		PSHA		
2A57	09		DEX		
2A58	5A		DECB		
2A59	26 F9		BNE	SQ4	
2A5B	86 02		LDAA	#2	
2A5D	BD 80 00		JSR	MATH	Y=X/IT
2A60	FE 18 39		LDX	XST	
2A63	C6 08		LDAB	#8	
2A65	A6 17	SQ5	LDAA	23D,X	
2A67	36		PSHA		
2A68	09		DEX		
2A69	5A		DECB		
2A6A	26 F9		BNE	SQ5	
2A6C	86 06		LDAA	#6	
2A6E	BD 80 00		JSR	MATH	Y=Y+IT
2A71	FE 18 39		LDX	XST	
2A74	6A 0F		DEC	15D,X	
2A76	C6 08		LDAB	#8	
2A78	32	SQ6	PULA		
2A79	A7 10		STAA	16D,X	
2A7B	08		INX		
2A7C	5A		DECB		
2A7D	26 F9		BNE	SQ6	
2A7F	7A 18 08		DEC	CTR	
2A82	26 CB		BNE	SQ3	
2A84	C6 08		LDAB	#8	
2A86	32	SQ7	PULA		
2A87	A7 10		STAA	16D,X	
2A89	08		INX		
2A8A	5A		DECB		

# METHOD OF INVARIANT MOMENTS

2AB8	26 F9		BNE	SQ7
2ABD	F6 18 4F		LDAB	STACK2
2A90	37		PSHB	
2A91	F6 18 4E		LDAB	STACK1
2A94	37		PSHB	
2A95	39		RTS	

\*  
\*  
\*

2A96	B6 18 30	TUNE	LDAA	UI1
2A99	B7 EE 11		STAA	\$EE11
2A9C	B6 18 31		LDAA	UI2
2A9F	B7 EE 10		STAA	\$EE10
2AA2	B6 18 32		LDAA	VJ1
2AA5	B7 EE 21		STAA	\$EE21
2AAB	B6 18 33		LDAA	VJ2
2AAB	B7 EE 20		STAA	\$EE20
2AAE	39		RTS	

\*  
\*  
\*

2AAF	B6 18 30	TUNE1	LDAA	UI1
2AB2	B7 EE 11		STAA	\$EE11
2AB5	B6 18 31		LDAA	UI2
2AB8	B7 EE 10		STAA	\$EE10
2ABB	B6 18 3B		LDAA	VVJ1
2ABE	B7 EE 21		STAA	\$EE21
2AC1	B6 18 3C		LDAA	VVJ2
2AC4	B7 EE 20		STAA	\$EE20
2AC7	39		RTS	

\*  
\*  
\*

2AC8	FF 18 00	IDEX	STX	TEMP1
2ACB	FB 18 01		ADDB	TEMP2
2ACE	B9 18 00		ADCA	TEMP1
2ADI	B7 18 00		STAA	TEMP1
2AD4	F7 18 01		STAB	TEMP2
2AD7	FE 18 00		LDX	TEMP1
2ADA	39		RTS	

\*  
\*  
\*

2ADB	B6 18 35	DELAY1	LDAA	DTIME
2ADE	4D	D1	TSTA	
2ADF	27 05		BEG	D2
2AE1	8D 04		BSR	DELAY
2AE3	4A		DECA	
2AE4	20 F8		BRA	D1
2AE6	39	D2	RTS	

\*  
\*  
\*

2AE7	CE 02 CA	DELAY	LDX	\$2CA
2AEA	09	LDY1	DEX	
2AEB	27 02		BEG	LDY2
2AED	20 FB		BRA	LDY1

# METHOD OF INVARIANT MOMENTS

2AEF	39		LDY2	RTS	
			*		
			*		
			*		
2AF0	7F 18 06		BI\$BCD	CLR	BCD1
2AF3	7F 18 07			CLR	BCD2
2AF6	86 10			LDAA	#16D
2AF8	B7 18 09			STAA	CTR1
2AFB	BF 18 04			STS	TEMP5
2AFE	8E 18 01			LDS	#TEMP3-1
2B01	86 08	LBIB1		LDAA	#8D
2B03	B7 18 08			STAA	CTR
2B06	33			PULB	
2B07	58	LBIB2		ASLB	
2B08	24 12			BCC	LBIB3
2B0A	86 01			LDAA	#1
2B0C	BB 18 07			ADDA	BCD2
2B0F	19			DAA	
2B10	B7 18 07			STAA	BCD2
2B13	B6 18 06			LDAA	BCD1
2B16	89 00			ADCA	#0
2B18	19			DAA	
2B19	B7 18 06			STAA	BCD1
2B1C	7A 18 09	LBIB3		DEC	CTR1
2B1F	27 1B			BEQ	LBIB4
2B21	B6 18 07			LDAA	BCD2
2B24	BB 18 07			ADDA	BCD2
2B27	19			DAA	
2B28	B7 18 07			STAA	BCD2
2B2B	B6 18 06			LDAA	BCD1
2B2E	B9 18 06			ADCA	BCD1
2B31	19			DAA	
2B32	B7 18 06			STAA	BCD1
2B35	7A 18 08			DEC	CTR
2B38	27 C7			BEQ	LBIB1
2B3A	20 CB			BRA	LBIB2
2B3C	BE 18 04	LBIB4		LDS	TEMP5
2B3F	39			RTS	
			*		
			*		
			*		
2B40	33		PULL8	PULB	
2B41	F7 18 4E			STAB	STACK1
2B44	33			PULB	
2B45	F7 18 4F			STAB	STACK2
2B48	C6 08			LDAB	#8
2B4A	32	P8		PULA	
2B4B	A7 00			STAA	0,X
2B4D	08			INX	
2B4E	5A			DECB	
2B4F	26 F9			BNE	P8
2B51	F6 18 4F			LDAB	STACK2
2B54	37			PSHB	
2B55	F6 18 4E			LDAB	STACK1
2B58	37			PSHB	
2B59	39			RTS	
			*		

# METHOD OF INVARIANT MOMENTS

* * *					
2B5A	33		PUSH88	PULB	
2B5B	F7 18 4E		STAB	STACK1	
2B5E	33		PULB		
2B5F	F7 18 4F		STAB	STACK2	
2B62	C6 08		LDAB	#8	
2B64	A6 07	P88	LDAA	7,X	
2B66	36		PSHA		
2B67	09		DEX		
2B68	5A		DECB		
2B69	26 F9		BNE	P88	
2B6B	F6 18 4F		LDAB	STACK2	
2B6E	37		PSHB		
2B6F	F6 18 4E		LDAB	STACK1	
2B72	37		PSHB		
2B73	39		RTS		
* * *					
2B74	F7 18 50	PUSH82	STAB	PUSHST	
2B77	33		PULB		
2B78	F7 18 4E		STAB	STACK1	
2B7B	33		PULB		
2B7C	F7 18 4F		STAB	STACK2	
2B7F	5F		CLRB		
2B80	37		PSHB		
2B81	37		PSHB		
2B82	37		PSHB		
2B83	37		PSHB		
2B84	37		PSHB		
2B85	37		PSHB		
2B86	F6 18 50		LDAB	PUSHST	
2B89	37		PSHB		
2B8A	36		PSHA		
2B8B	86 07		LDAA	#7	
2B8D	BD 80 00		JSR	MATH	NFLPTN
2B90	F6 18 4F		LDAB	STACK2	
2B93	37		PSHB		
2B94	F6 18 4E		LDAB	STACK1	
2B97	37		PSHB		
2B98	39		RTS		
* * *					
2B99	33	PULL4	PULB		
2B9A	F7 18 4E		STAB	STACK1	
2B9D	33		PULB		
2B9E	F7 18 4F		STAB	STACK2	
2BA1	4D		TSTA		
2BA2	27 06		BEQ	LP42	
2BA4	08	LP41	INX		
2BA5	4A		DECA		
2BA6	27 02		BEQ	LP42	
2BA8	20 FA		BRA	LP41	
2BAA	32	LP42	PULA		
2BAB	A7 00		STAA	0,X	



# METHOD OF INVARIANT MOMENTS

2BAD	32		PULA	
2BAE	A7 01		STAA	1,X
2BB0	32		PULA	
2BB1	A7 02		STAA	2,X
2BB3	32		PULA	
2BB4	A7 03		STAA	3,X
2BB6	F6 18 4F		LDAB	STACK2
2BB9	37		PSHB	
2BBA	F6 18 4E		LDAB	STACK1
2BBD	37		PSHB	
2BBE	39		RTS	

\*  
\*  
\*

2BBF	33	PUSH44	PULB	
2BC0	F7 18 4E		STAB	STACK1
2BC3	33		PULB	
2BC4	F7 18 4F		STAB	STACK2
2BC7	4D		TSTA	
2BC8	27 06		BEQ	LP442
2BCA	08	LP441	INX	
2BCB	4A		DECA	
2BCC	27 02		BEQ	LP442
2BCE	20 FA		BRA	LP441
2BD0	A6 03	LP442	LDAA	3,X
2BD2	36		PSHA	
2BD3	A6 02		LDAA	2,X
2BD5	36		PSHA	
2BD6	A6 01		LDAA	1,X
2BD8	36		PSHA	
2BD9	A6 00		LDAA	0,X
2BDB	36		PSHA	
2BDC	F6 18 4F		LDAB	STACK2
2BDF	37		PSHB	
2BE0	F6 18 4E		LDAB	STACK1
2BE3	37		PSHB	
2BE4	39		RTS	

\*  
\*  
\*

2BE5	33	PUSH41	PULB	
2BE6	F7 18 4E		STAB	STACK1
2BE9	33		PULB	
2BEA	F7 18 4F		STAB	STACK2
2BED	36		PSHA	
2BEE	48		ASLA	
2BEF	24 04		BCC	LP411
2BF1	86 FF		LDAA	##FF
2BF3	20 01		BRA	LP412
2BF5	4F	LP411	CLRA	
2BF6	36	LP412	PSHA	
2BF7	36		PSHA	
2BF8	36		PSHA	
2BF9	F6 18 4F		LDAB	STACK2
2BFC	37		PSHB	
2BFD	F6 18 4E		LDAB	STACK1
2C00	37		PSHB	

# METHOD OF INVARIANT MOMENTS

2C62	A7	27		STAA	\$27,X
2C64	6A	08		DEC	8,X
2C66	20	CE		BRA	RD1
2C68	A6	3E	RD2	LDAA	\$3E,X
2C6A	A7	08		STAA	8,X
2C6C	6D	08	RD3	TST	8,X
2C6E	27	22		BEQ	RD4
2C70	A6	24		LDAA	\$24,X
2C72	47			ASRA	
2C73	A7	24		STAA	\$24,X
2C75	A6	25		LDAA	\$25,X
2C77	46			RORA	
2C78	A7	25		STAA	\$25,X
2C7A	A6	26		LDAA	\$26,X
2C7C	46			RORA	
2C7D	A7	26		STAA	\$26,X
2C7F	A6	27		LDAA	\$27,X
2C81	47			ASRA	
2C82	A7	27		STAA	\$27,X
2C84	A6	28		LDAA	\$28,X
2C86	46			RORA	
2C87	A7	28		STAA	\$28,X
2C89	A6	29		LDAA	\$29,X
2C8B	46			RORA	
2C8C	A7	29		STAA	\$29,X
2C8E	6A	08		DEC	8,X
2C90	20	DA		BRA	RD3
2C92	FE	18 25	RD4	LDX	COR2
2C95	FF	18 24		STX	COR1
2C98	FE	18 28		LDX	COR4+1
2C9B	FF	18 26		STX	COR3
2C9E	39			RTS	

* * *					
2C9F	86	03	READSN	LDAA	#3
2CA1	BD	33 B8		JSR	MUXSEL
2CA4	CE	18 00		LDX	#TEMP1
2CA7	BD	33 0F		JSR	SYNDET
2CAA	7D	18 00		TST	TEMP1
2CAD	8D	0C		BSR	SGNSET
2CAF	87	18 A1		STAA	SIGNI
2CB2	7D	18 02		TST	TEMP3
2CB5	8D	04		BSR	SGNSET
2CB7	87	18 A2		STAA	SIGNQ
2CBA	39			RTS	

* * *					
2CBB	2A	03	SGNSET	BPL	SETPOS
2CBD	86	FF		LDAA	##FF
2CBF	39			RTS	
2CC0	4F		SETPOS	CLRA	
2CC1	39			RTS	

# METHOD OF INVARIANT MOMENTS

2CC2			C01	EQU	*
2CC2	40	00	98	FCB	\$40,0,\$98,\$33
2CC6	00	00	0B	FCB	0,0,\$B,\$95
2CCA	00	00	FB	FCB	0,0,\$FB,\$CD
2CCE	00	00	02	FCB	0,0,2,\$2C
2CD2	00	00	FE	FCB	0,0,\$FE,\$AB
2CD6	00	00	00	FCB	0,0,0,\$EA
2CDA	00	00	FF	FCB	0,0,\$FF,\$55
2CDE	00	00	00	FCB	0,0,0,\$85
2CE2	00	00	FF	FCB	0,0,\$FF,\$96
2CE6	00	00	00	FCB	0,0,0,\$59
2CEA	00	00	FF	FCB	0,0,\$FF,\$B5
2CEE	00	00	00	FCB	0,0,0,\$42
2CF2	00	00	FF	FCB	0,0,\$FF,\$C6
2CF6	00	00	00	FCB	0,0,0,\$36
2CFA	00	00	FF	FCB	0,0,\$FF,\$CF
2CFE	00	00	00	FCB	0,0,0,\$3D
2D02	00	00	FF	FCB	0,0,\$FF,\$D5
2D06	00	00	00	FCB	0,0,0,\$2B
2D0A	00	00	FF	FCB	0,0,\$FF,\$D7
2D0E	00	00	00	FCB	0,0,0,\$29
2D12			C23	EQU	*
2D12	40	00	90	FCB	\$40,0,\$90,\$44
2D16	BD	F4	1D	FCB	\$BD,\$F4,\$1D,\$E
2D1A	00	00	0A	FCB	0,0,\$A,\$95
2D1E	02	72	FC	FCB	2,\$72,\$FC,\$51
2D22	00	00	FE	FCB	0,0,\$FE,\$9F
2D26	FF	7A	00	FCB	\$FF,\$7A,0,\$C5
2D2A	00	00	00	FCB	0,0,0,\$8B
2D2E	00	31	FF	FCB	0,\$31,\$FF,\$AB
2D32	00	00	FF	FCB	0,0,\$FF,\$CB
2D36	FF	EA	00	FCB	\$FF,\$EA,0,\$27
2D3A	00	00	00	FCB	0,0,0,\$1F
2D3E	00	0C	FF	FCB	0,\$C,\$FF,\$EA
2D42	00	00	FF	FCB	0,0,\$FF,\$F1
2D46	FF	FA	00	FCB	\$FF,\$FA,0,\$C
2D4A	00	00	00	FCB	0,0,0,\$A
2D4E	00	04	FF	FCB	0,4,\$FF,\$F9
2D52	00	00	FF	FCB	0,0,\$FF,\$FC
2D56	FF	FF	00	FCB	\$FF,\$FF,0,3
2D5A	00	00	00	FCB	0,0,0,2
2D5E	00	01	00	FCB	0,1,0,0
2D62			C45	EQU	*
2D62	40	00	8E	FCB	\$40,0,\$8E,\$71
2D66	B4	EF	21	FCB	\$B4,\$EF,\$21,\$B5
2D6A	00	00	10	FCB	0,0,\$10,\$2A
2D6E	0D	5A	F7	FCB	\$D,\$5A,\$F7,\$8E
2D72	00	00	FB	FCB	0,0,\$FB,\$70
2D76	FC	F1	02	FCB	\$FC,\$F1,2,\$98
2D7A	00	00	01	FCB	0,0,1,\$96
2D7E	01	21	FE	FCB	1,\$21,\$FE,\$F5
2D82	00	00	FF	FCB	0,0,\$FF,\$47
2D86	FF	7A	00	FCB	\$FF,\$7A,0,\$84
2D8A	00	00	00	FCB	0,0,0,\$61
2D8E	00	48	FF	FCB	0,\$48,\$FF,\$B8
2D92	00	00	FF	FCB	0,0,\$FF,\$CA
2D96	FF	DB	00	FCB	\$FF,\$DB,0,\$2A

# METHOD OF INVARIANT MOMENTS

2D9A	00 00 00		FCB	0,0,0,\$20
2D9E	00 18 FF		FCB	0,\$18,\$FF,\$E9
2DA2	00 00 FF		FCB	0,0,\$FF,\$F0
2DA6	FF F5 00		FCB	\$FF,\$F5,0,\$C
2DAA	00 00 00		FCB	0,0,0,7
2DAE	00 04 FF		FCB	0,4,\$FF,\$FF
2DB2		C67	EQU	*
2DB2	40 00 8D		FCB	\$40,0,\$8D,\$BF
2DB6	B2 03 23		FCB	\$B2,3,\$23,\$98
2DBA	00 00 12		FCB	0,0,\$12,\$BC
2DBE	12 A5 F4		FCB	\$12,\$A5,\$F4,\$E2
2DC2	00 00 F9		FCB	0,0,\$F9,\$18
2DC6	F9 95 04		FCB	\$F9,\$95,4,\$68
2DCA	00 00 02		FCB	0,0,2,\$E9
2DCE	02 96 FE		FCB	2,\$96,\$FE,\$5
2DD2	00 00 FE		FCB	0,0,\$FE,\$9B
2DD6	FE C1 01		FCB	\$FE,\$C1,1,4
2DDA	00 00 00		FCB	0,0,0,\$C1
2DDE	00 AE FF		FCB	0,\$AE,\$FF,\$EF
2DE2	00 00 FF		FCB	0,0,\$FF,\$92
2DE6	FF 9D 00		FCB	\$FF,\$9D,0,\$55
2DEA	00 00 00		FCB	0,0,0,\$41
2DEE	00 3A FF		FCB	0,\$3A,\$FF,\$D0
2DF2	00 00 FF		FCB	0,0,\$FF,\$DD
2DF6	FF E2 00		FCB	\$FF,\$E2,0,\$18
2DFA	00 00 00		FCB	0,0,0,\$E
2DFE	00 0A FF		FCB	0,\$A,\$FF,\$FC
2E02	CE 40 00	DUMMY	LDX	\$4000
2E05	FF 18 24		STX	COR1
2E08	CE 00 00		LDX	\$0
2E0B	FF 18 24		STX	COR3
2E0E	39		RTS	
2E0F			END	

STATEMENTS =832

FREE BYTES =16585

NO ERRORS DETECTED

## APPENDIX C - Fortran Code Listings

This Appendix consists of listings of Fortran programs which can be used to compute the coefficients  $c_k(p)$  which are defined by equations (104), (105) and (106). These coefficients are used to compute image moments from the image Fourier transform. The program PEVEN computes the coefficients for the case that  $p$  is even. The program PODD computes the coefficients for the case that  $p$  is odd.

This Appendix also contains listings of the programs EVENER and ODDER. These programs were used to obtain the data which is presented in Figure 26 and Tables 3 and 4. EVENER evaluates equation (137) for the case of  $p$  even. ODDER evaluates the same equation for the case of  $p$  odd.

```

/   PEVEN
    PI=3.1415927
    TYPE 10
10   FORMAT(' COMPUTES CK(P); P EVEN')
    TYPE 20
20   FORMAT(' ')
    TYPE 30
30   FORMAT(' INPUT N/2, P(MAX)')
    TYPE 20
    ACCEPT 40,N
    ACCEPT 40,NPM
40   FORMAT(I3)
    N=N+N
    TYPE 20
    TYPE 50
50   FORMAT(' COEFFICIENTS (REAL)')
    TYPE 20
    TYPE 60
60   FORMAT(' K      P      CK(P)      OCTAL')
    DO 105 NP=0,NPM,2
    DO 100 K=0,N,1
    C=0.
    IF(NP.EQ.0) C=.5
    DO 90 L=1,N,1
    X=2.*FLOAT(L)/FLOAT(N)
    IF(X.GT.1.) GO TO 70
    FX=X**FLOAT(NP)
    GO TO 80
70   FX=2-(X-2)**FLOAT(NP)
80   ARG=FLOAT(K)*PI*X/2.
    IF(L.EQ.N) C=C-.5*FX*COS(ARG)
90   C=C+FX*COS(ARG)
    C=C/FLOAT(N)
    NC=.5+32768.*C
    IF(NC.LT.0) NC=65536+NC
100  TYPE 110,K,NP,C,NC
105  TYPE 20
110  FORMAT(I3,I5,' ',F11.7,' ',06)
    STOP
    END

```

\*\*

```

/  PDDD
    PI=3.1415927
    TYPE 10
10  FORMAT(' COMPUTES CK(P); P ODD')
    TYPE 20
20  FORMAT(' ')
    TYPE 30
30  FORMAT(' INPUT N/4, P(MAX)')
    TYPE 20
    ACCEPT 40,N
    ACCEPT 40,NPM
40  FORMAT (I3)
    N=N+N+N+N
    TYPE 20
    TYPE 50
50  FORMAT(' COEFFICIENTS (IMAGINARY)')
    TYPE 20
    TYPE 60
60  FORMAT(' K      P      CK(P)      OCTAL')
    DO 140 NP=1,NPM,2
    DO 130 K=1,N,1
    C=0.
    DO 120 L=1,N,1
    X=4.*FLOAT(L)/FLOAT(N)
    IF(X.GT.1.) GO TO 70
    FX=X**FLOAT(NP)
    GO TO 110
70  IF(X.GT.2.) GO TO 90
80  FX=2.+(X-2.)*FLOAT(NP)
    GO TO 110
90  IF(X.GT.3.) GO TO 100
    FX=2.-(X-2.)*FLOAT(NP)
    GO TO 110
100 FX=-(X-4.)*FLOAT(NP)
110 ARG=FLOAT(K)*PI*X/4.
    IF(L.EQ.N) C=C-.5*FX*SIN(ARG)
120 C=C+FX*SIN(ARG)
    C=-C/FLOAT(N)
    NC=.5+32768.*C
    IF(NC.LT.0) NC=65536+NC
130 TYPE 150, K,NP,C,NC
140 TYPE 20
150 FORMAT (I3,I5,' ',F11.7,' ',06)
    STOP
    END

```

\*\*

```

/  EVENER

    DIMENSION CFF(50)
    PI=3.1415927
    TYPE 10
10  FORMAT(' COMPUTES ER(P,X); P EVEN')
15  TYPE 20
20  FORMAT(' ')
    TYPE 30
30  FORMAT(' INPUT N/2, P')
    TYPE 20
    ACCEPT 40,N
    ACCEPT 40,NP
40  FORMAT(I3)
    N=N+N
    TYPE 20
    DO 100 K=0,N,1
    C=0.
    IF(NP.EQ.0) C=.5
    DO 90 L=1,N,1
    X=2.*FLOAT(L)/FLOAT(N)
    IF(X.GT.1.) GO TO 70
    FX=X**FLOAT(NP)
    GO TO 80
70  FX=2-(X-2)**FLOAT(NP)
80  ARG=FLOAT(K)*PI*X/2.
    IF(L.EQ.N) C=C-.5*FX*COS(ARG)
90  C=C+FX*COS(ARG)
    C=C/FLOAT(N)
100 CFF(K+1)=C
    TYPE 110
110 FORMAT(' INPUT MESH DIVISOR=M')
    TYPE 20
    ACCEPT 40,M
    TYPE 20
    TYPE 120
120 FORMAT(' Q      ER(P,2*Q/MN)')
    NQS=.5*FLOAT(N*M)
    DO 140 NQ=0,NQS,1
    X=FLOAT(2*NQ)/FLOAT(M*N)
    ARG=PI*X/2.
    ER=0.
    DO 130 K=1,N,1
    ARGK=ARG*FLOAT(K)
130  ER=ER-CFF(K+1)*COS(ARGK)
    ER=ER-.5*(CFF(1)-CFF(N+1)*COS(ARGK))
    ER=2.*ER+X**FLOAT(NP)
140  TYPE 150,NQ,ER
150  FORMAT(I4,' ',F11.7)
    TYPE 160
160  FORMAT(' TO CONTINUE TYPE 1')
    TYPE 20
    ACCEPT 40,N
    IF(N-1)170,15,170
170  STOP
    END
**

```



```

/  ODDER

    DIMENSION CFF(50)
    PI=3.1415927
    TYPE 10
10  FORMAT(' COMPUTES ER(P,X); P ODD')
15  TYPE 20
20  FORMAT(' ')
    TYPE 30
30  FORMAT(' INPUT N/4, P')
    TYPE 20
    ACCEPT 40,N
    ACCEPT 40,NP
40  FORMAT (I3)
    N=N+N+N+N
    TYPE 20
    DO 130 K=1,N,1
    C=0.
    DO 120 L=1,N,1
    X=4.*FLOAT(L)/FLOAT(N)
    IF(X.GT.1.) GO TO 70
    FX=X**FLOAT(NP)
    GO TO 110
70  IF(X.GT.2.) GO TO 90
80  FX=2.*(X-2.）**FLOAT(NP)
    GO TO 110
90  IF(X.GT.3.) GO TO 100
    FX=2.-(X-2.）**FLOAT(NP)
    GO TO 110
100  FX=-(X-4.）**FLOAT(NP)
110  ARG=FLOAT(K)*PI*X/4.
    IF(L.EQ.N) C=C-.5*FX*SIN(ARG)
120  C=C+FX*SIN(ARG)
    C=C/FLOAT(N)
130  CFF(K+1)=C
    TYPE 140
140  FORMAT(' INPUT MESH DIVISOR=M')
    TYPE 20
    ACCEPT 40,M
    TYPE 20
    TYPE 150
150  FORMAT(' Q          ER(P,2*Q/MN)')
    NQS=.25*FLOAT(N*M)
    DO 170 NQ=0,NQS,1
    X=FLOAT(4*NQ)/FLOAT(M*N)
    ARG=PI*X/4.
    ER=0.
    DO 160 K=1,N,1
    ARGK=ARG*FLOAT(K)
160  ER=ER-CFF(K+1)*SIN(ARGK)
    ER=2.*ER+X**FLOAT(NP)
170  TYPE 180,NQ,ER
180  FORMAT(I4,' ',F11.7)
    TYPE 190
190  FORMAT(' TO CONTINUE TYPE 1')
    TYPE 20
    ACCEPT 40,N
    IF(N-1) 200,15,200
200  STOP
    END

```

Distribution:

W26HAJ

2 copies

U.S. Army Engineer Topographic Laboratories  
Research Institute  
Fort Belvoir, VA 22060  
Attn: Joseph F. Hannigan

W26HAJ/STINFO

14 copies

U.S. Army Engineer Topographic Laboratories  
Attn: ETL - STINFO  
Fort Belvoir, VA 22060

END

DATE  
FILMED

03-82

DTIC